

### Raytheon

# CLOUD EFFECTIVE PARTICLE SIZE AND CLOUD OPTICAL THICKNESS

## VISIBLE/INFRARED IMAGER/RADIOMETER SUITE ALGORITHM THEORETICAL BASIS DOCUMENT

Version 4: May 2001

Szu-Cheng Ou Kuo-Nan Liou Yoshihide Takano Glenn J. Higgins North Larsen Richard Slonaker

RAYTHEON SYSTEMS COMPANY Information Technology and Scientific Services 4400 Forbes Boulevard Lanham, MD 20706

SBRS Document #: Y2393

EDRs: CLOUD EFFECTIVE PARTICLE SIZE

AND CLOUD OPTICAL THICKNESS

Doc No: Y2393

Version: 4

Revision: 0

	Function	Name	Signature	Date
PREPARED BY	EDR DEVELOPER	S. OU		5/3/01
APPROVED BY	RELEVENT LEAD	R. SLONAKER		5/13/01
APPROVED BY	CHIEF SCIENTIST	S. MILLER		5/14/01
RELEASED BY	ALGORITHM LEAD	P. KEALY		5/15/01

#### **TABLE OF CONTENTS**

			<u>Page</u>
LIST	Γ OF FI	GURES	iv
LIST	ΓOF TA	ABLES	viii
GLC	SSAR	Y OF ACRONYMS	ix
ABS	STRAC	Γ	xi
1.0	INTRO	DDUCTION	1
	1.1	PURPOSE	1
	1.2	SCOPE	1
	1.3	VIIRS DOCUMENTS	1
	1.4	REVISIONS	1
2.0	EXPE	RIMENT OVERVIEW	3
	2.1	OBJECTIVES OF CLOUD OPTICAL THICKNESS AND EFFECTIVE PARTICLE SIZE RETRIEVALS	3
	2.2	INSTRUMENT CHARACTERISTICS	5
	2.3	RETRIEVAL STRATEGY	7
3.0	ALGO	PRITHM DESCRIPTION	9
	3.1	PROCESSING OUTLINE	9
	3.2	ALGORITHM INPUT	10
		3.2.1 VIIRS Data	10
		3.2.1.1 Calibrated Brightness Temperatures and Reflectances	
		3.2.1.2 Geometric Parameters	10
		3.2.1.3 Exo-Atmospheric Solar Spectral Irradiances	
		3.2.1.4 Cloud Mask Product	11
		3.2.1.5 VIIRS Band Parameters	
		3.2.1.6 Cloud Top Height	
		3.2.1.7 Surface Type	11

		3.2.1.8 Surface Tem	perature	11
	3.2.2	Non-VIIRS Data		11
			Sounding	
3.3	THEOR	ETICAL DESCRIPTION	OF OPTICAL DEPTH AND EFFECTIVE	
	PARTIC	LE SIZE RETRIEVALS	)	12
	3.3.1	Physics of the Problem		12
	3.3.1		al and Radiative Properties of Cirrus Clouds	
			Optical Properties	
			erties for Rayleigh Scattering	
			tering	
			Distribution Approach	
		3.3.1.6 Single-Scatte	ering Properties for Combined Cloud and	
		Aerosol Part	icles, Rayleigh Molecules, and Gases	22
		3.3.1.7 Radiative Tr	ansfer Model	23
	3.3.2	Mathematical Description	on of the Algorithms	37
		3.3.2.1 Retrieval of	Cirrus Cloud Parameters	37
		3.3.2.2 Retrieval of	Water Cloud Parameters	46
	3.3.3	Archived Algorithm Ou	tput	52
	3.3.4	Variance and Uncertain	ty Estimates	52
2.4	A L COD			
3.4			ΓUDIES ON RETRIEVALS OF OPTICAL	5/
	ITICK	(E33		34
	3.4.1	SNR Tests		55
		3.4.1.1 Daytime Res	ults	56
		C	esults	
		3.4.1.3 Recommend	ations	84
	3.4.2	Error Budget Studies		85
			pecific Errors in the Cloud Module	
			urces	
			Sources	
		3.4.2.4 Specified an	d Predicted Performances	86
3.5	ALGOR	ITHM SENSITIVITY S	ΓUDIES ON RETRIEVALS OF EFFECTIVE	
	PARTIC	LE SIZE		87
	3.5.1	SNR Tests		88
	3.3.1		ults	
		-	esults	
			ations	
	3.5.2			
	5.5.4	•	pecific Errors in the Cloud Module	
			urces	
			Sources	
			for Cloud Effective Particle Size	

	3.6	PRACT	TICAL CONSIDERATIONS	113
		3.6.1	Numerical Computation Considerations	113
		3.6.2	Programming and Procedural Considerations	114
		3.6.3	Configuration of Retrievals	114
		3.6.4	Quality Assessment and Diagnostics	114
		3.6.5	Exception Handling	115
	3.7	ALGO	RITHM VALIDATION	115
	3.8	ALGO	RITHM DEVELOPMENT SCHEDULE	116
4.0	ASSU	MPTION	NS AND LIMITATIONS	117
	4.1	ASSUM	MPTIONS	117
	4.2	LIMITA	ATIONS	117
5.0	REFE	RENCES	S	118
App	endix A	. SRD RI	EQUIREMENTS	122
App	endix B	B. ESTIM	ATING MEAN CIRRUS CLOUD TEMPERATURE	124
App	endix C	. ITERA	TIVE MATCHING OF REFLECTANCE PAIRS	126
App	endix D	.ESTIM	ATING MEAN WATER CLOUD TEMPERATURE	128

#### **LIST OF FIGURES**

	<u> 1</u>	Page
Figure 1.	Spectral characteristics of the seven VIIRS solar channels, centered at 0.672, 0.859, 1.24, 1.378, 1.6, 2.25, and 3.7 $\mu$ m, used for retrievals of optical thickness and effective particle size. The atmospheric zenith transmittances are calculated from the LBLE model for three levels at 10 km, 5 km and surface based on the US Standard Atmosphere at 32° solar zenith angle.	7
Figure 2.	A high-level flow chart describing the general retrieval strategy for cloud optical thickness and effective particle size.	8
Figure 3. A	A high-level flow diagram depicting the processing outline of the data flow of the VIIRS channel radiance in the development of VIIRS cloud EDR retrieval algorithms.	9
Figure 4.	The six discrete ice crystal size distributions used in generating radiances and algorithm sensitivity studies	14
Figure 5.	Phase functions for the six size distributions and for the five VIIRS cloud retrieval channels	18
Figure 6a.	Single scattering albedos for the six size distributions and for the five VIIRS cloud retrieval channels.	19
Figure 6b.	. Asymmetry factors for the six size distributions and for the five VIIRS cloud retrieval channels.	19
Figure 7.	Flow diagram of the computation of TOA radiance for the application to cloud remote sensing using VIIRS channels.	27
Figure 8.	Sun satellite geometry.	29
Figure 9.	The contour plot of the 0.645 $\mu m$ reflectance and radiance for various combinations of cirrus cloud optical depths and mean effective sizes and for high-sun angle ( $\theta_0 = 32^\circ$ ) sensor at nadir angle.	33
Figure 10.	. The contour plot of the 3.75 $\mu$ m total (solar + thermal) reflectance and radiance for the same combination of cirrus cloud optical depths and mean effective sizes and for high-sun angle ( $\theta_0 = 32^\circ$ ) sensor at nadir angle.	34
Figure 11.	. The contour plot of the 0.645 and 3.75 μm radiance percentage differences between LBLE and MODTRAN-DISORT	35
Figure 12.	. The contour plots of the 10.8 µm MODTRAN-DISORT radiance and the differences between LBLE and MODTRAN-DISORT radiance.	36
Figure 13.	. Flow chart for the cirrus cloud retrieval program	37
Figure 14.	. Flow chart for nighttime IR cirrus retrieval algorithm.	42

Figure 15.	MAS data taken from FIRE-II IFO over both land and water surfaces. The plane-parallel radiative transfer calculations were performed for the six ice crystal size distributions presented in section 3.3.1.1 and for two water clouds with mean effective radii of 4 and 8 μm. The optical depth ranges from 0.5 to 64	44
Figure 16.	Flow chart for cirrus solar retrieval program.	45
Figure 17.	Flowchart for the daytime IR cirrus cloud retrieval program.	46
Figure 18.	Flow chart for the water cloud retrieval program.	47
Figure 19.	Flow chart for the water cloud solar retrieval program.	48
Figure 20.	Flow chart for nighttime IR water cloud retrieval algorithm.	52
Figure 21.	The correlation between the VIIRS $0.645$ and $1.6~\mu m$ reflectances and between the $0.645$ and $2.13~\mu m$ reflectances for cirrus clouds in US Standard Atmosphere with the sensor pointing at nadir.	58
Figure 22.	Accuracy and Precision of retrieved optical depths from the cirrus solar algorithm for US Standard Atmosphere, sensor at nadir, based on the 0.645 – 1.61 µm correlation.	59
Figure 23.	Accuracy and Precision of retrieved optical depths from the cirrus solar algorithm for US Standard Atmosphere, sensor at nadir, based on the 0.645 – 2.13 µm correlation.	61
Figure 24.	Display of the 0.645-1.6 µm and 0.645-2.13 µm reflectance correlations for cirrus cloud in US Standard Atmosphere, sensor at off-nadir	62
Figure 25.	Accuracy and Precision of retrieved optical depths from the cirrus solar algorithm for US Standard Atmosphere, sensor at off-nadir	64
Figure 26.	Display of the 0.645-1.6 µm and 0.645-2.13 µm reflectance correlations for cirrus cloud in US Standard Atmosphere, sensor at edge-of-scan.	65
Figure 27.	Accuracy and Precision of retrieved optical depths from the solar algorithm for cirrus cloud in US Standard Atmosphere with sensor pointing at edge-of-scan	66
Figure 28.	The 0.645-1.61 µm and 0.645-2.13 µm reflectance correlations for cirrus clouds in the Tropical Atmosphere, sensor at nadir.	68
Figure 29.	Accuracy and Precision of retrieved optical depths from the solar algorithm for cirrus cloud in Tropical Atmosphere with sensor pointing at nadir	69
Figure 30.	The 0.645-1.6 µm reflectance correlations for cirrus cloud in Subarctic Atmosphere, sensor at nadir.	70
Figure 31.	Accuracy and Precision of retrieved optical depths from the solar algorithm for cirrus cloud in Subarctic Atmosphere with sensor pointing at nadir	71

Figure 32.	The 0.645-1.6 µm reflectance correlations for cirrus cloud in Desert Atmosphere, sensor at nadir.	73
Figure 33.	Accuracy and Precision of retrieved optical depths from the solar algorithm for cirrus cloud in Desert Atmosphere with sensor pointing at nadir	74
Figure 34.	The 0.645-1.6 µm and 0.645-2.13 µm reflectance correlations for cirrus cloud in US Standard Atmosphere, sensor at nadir.	76
Figure 35.	Accuracy and Precision of retrieved optical depths from the solar algorithm for water cloud in US Standard Atmosphere with sensor pointing at nadir.	77
Figure 36.	Accuracy and Precision of retrieved optical depths from the IR algorithm for cirrus cloud in US Standard Atmosphere with sensor pointing at nadir	79
Figure 37.	Accuracy and Precision of retrieved optical depths from the IR algorithm for cirrus cloud in US Standard Atmosphere with sensor pointing at off-nadir	80
Figure 38.	Accuracy and Precision of retrieved optical depths from the IR algorithm for cirrus cloud in US Standard Atmosphere with sensor pointing at edge-of-scan	81
Figure 39.	Accuracy and Precision of retrieved optical depths from the IR algorithm for cirrus cloud in Tropical Atmosphere with sensor pointing at edge-of-scan.	83
Figure 40.	Accuracy and Precision of retrieved optical depths from the IR algorithm for water cloud in US Standard Atmosphere with sensor pointing at nadir for nonoise retrievals.	84
Figure 41.	Accuracy and Precision of retrieved effective particle size from the cirrus solar algorithm for US Standard Atmosphere, sensor at nadir, based on the 0.645-1.61 µm correlation.	90
Figure 42.	Accuracy and Precision of retrieved effective particle size from the cirrus solar algorithm for US Standard Atmosphere, sensor at nadir, based on the 0.645-2.13 µm correlation.	91
Figure 43.	Accuracy and Precision of retrieved effective particle size from the cirrus solar algorithm for US Standard Atmosphere, sensor at off-nadir, based on the 0.645-1.61 µm correlation.	92
Figure 44.	Accuracy and Precision of retrieved effective particle size from the cirrus solar algorithm for US Standard Atmosphere, sensor at edge-of-scan, based on the 0.645-1.61 µm correlation	93
Figure 45.	Accuracy and Precision of retrieved effective particle size from the cirrus solar algorithm for Tropical Atmosphere, sensor at nadir, based on the $0.645$ - $1.61~\mu m$ correlation.	95
Figure 46.	Accuracy and Precision of retrieved effective particle size from the cirrus solar algorithm for Subarctic Atmosphere, sensor at nadir, based on the 1.2-1.61 µm correlation.	96

Figure 47.	Accuracy and Precision of retrieved effective particle size from the cirrus solar algorithm for Desert Atmosphere, sensor at nadir, based on the 0.645-1.61 μm correlation.	.97
Figure 48.	Accuracy and Precision of retrieved effective particle size from the water cloud solar algorithm for US Standard Atmosphere, sensor at nadir, based on the 0.645-1.61 µm correlation.	.98
Figure 49.	Accuracy of retrieved effective particle size from the cirrus cloud IR algorithm for US Standard Atmosphere, sensor at nadir for $D_e$ = 23.9 $\mu$ m, 41.5 $\mu$ m, and 93 $\mu$ m.	103
Figure 50.	Precision of retrieved effective particle size from the cirrus cloud IR algorithm for US Standard Atmosphere, sensor at nadir for $D_e = 23.9 \mu\text{m}$ , and 41.5 $\mu\text{m}$	104
Figure 51.	Accuracy of retrieved effective particle size from the cirrus cloud IR algorithm for US Standard Atmosphere, sensor at off-nadir for $D_e$ = 23.9 $\mu$ m, 41.5 $\mu$ m, and 93 $\mu$ m.	105
Figure 52.	Precision of retrieved effective particle size from the cirrus cloud IR algorithm for US Standard Atmosphere, sensor at off-nadir for $D_e$ = 23.9 $\mu$ m, and 41.5 $\mu$ m	106
Figure 53.	Accuracy of retrieved effective particle size from the cirrus cloud IR algorithm for US Standard Atmosphere, sensor at edge-of-scan for $D_e$ = 23.9 $\mu$ m, 41.5 $\mu$ m, and 93 $\mu$ m.	107
Figure 54.	Same as Figure 63, except for Precision.	108
Figure 55.	Accuracy of retrieval effective particle size for the cirrus cloud IR algorithm for tropical atmosphere, sensor at nadir for $D_e = 41.5 \mu\text{m}$ , 23.9 $\mu\text{m}$ , and 93 $\mu\text{m}$	109
Figure 56.	Precision of retrieved effective particle size from the cirrus cloud IR algorithm for Tropical Atmosphere, sensor at nadir for $D_e = 41.5 \text{ um}$ .	110

#### **LIST OF TABLES**

		<u>Page</u>
Table 1.	VIIRS Baseline Bands and Bandwidths for the Cloud EDRs	6
Table 2.	Mean Effective Size of the Six Ice Crystal Size Distributions Employed in the Present Work	14
Table 3.	Comparison of the 0.645 µmclear and haze radiance from the LBLE model (UCLA) and the MODTRAN/MODTRAN-DISORT model for various sunsensor geometries, including high-sun, low-sun, nadir-sensor, and off-nadir-sensor configurations.	30
Table 4.	Comparison of the 3.75 $\mu m$ clear and haze radiance from the LBLE model (UCLA) and the MODTRAN/MODTRAN-DISORT model for various sunsensor geometries, including high-sun, low-sun, and nadir-sensor configurations.	31
Table 5.	Comparison of the 10.8 µm clear and haze radiance from the LBLE model (UCLA) and the MODTRAN/MODTRAN-DISORT model for various sensor geometries, including nadir-sensor and off-nadir-sensor configurations	31
Table 6.	Input Parameters for Each Scenario	56
Table A	1. Thresholds and objectives of the cloud optical depth EDR parameters	. 122
Table A2	2. Thresholds and objectives of the cloud effective particle size EDR parameters	. 123

#### **GLOSSARY OF ACRONYMS**

ATBD Algorithm Theoretical Basis Document

AVHRR Advanced Very-High-Resolution Radiometer

Ci Cirrus

CLASS Cross-chain LORAN Atmospheric Sounding System

COT Cloud Optical Thickness

Cs Cirrostratus

CSSM Cloud Scene Simulation Model
DAAC Distributed Active Archive Center

DISORT Discrete Ordinate Radiative Transfer Model
DMSP Defense Meteorological Satellite Program

DOC Department of Commerce DoD Department of Defense

EDR Environmental Data Record

EPS Effective Particle Size

FASCOD Fast Atmosphere Signature Code

FDTD Finite-Difference Time Domain Method

FIRE First ISCCP Regional Experiment

HCS Horizontal Cell Size

HITRAN High-Resolution Transmission Molecular Absorption Database

IFO Intensive Field Observation
IPT Integrated Project Team

IR Infrared

ISCCP International Satellite Cloud Climatology Project

IWC Ice Water Content

LBLE Line-by-Line Equivalent Radiative Transfer Model

LOWTRAN Low-Resolution Transmission Model

MAS MODIS Airborne Simulator

MCR Multi-spectral Cloud Radiometer
MD MODTRAN-DISORT program

MODIS Moderate Resolution Imaging Spectroradiometer

MODTRAN Moderate-Resolution Atmospheric Transmittance and Radiance Model

NCAR National Center for Atmospheric Research

NOAA National Oceanic and Atmospheric Administration

NPOESS National Polar-orbiting Operational Environmental Satellite System

POES Polar-orbiting Operational Environmental Satellite

SNR Signal-Noise Ratio

SRD Sensor Requirement Document

SUCCESS Subsonic Aircraft Contrail and Cloud Effects Special Study

TBD To be determined by contractor

TBR To be resolved

VIIRS Visible/Infrared Imager/Radiometer Suite

#### **ABSTRACT**

This Algorithm Theoretical Basis Document (ATBD) describes the algorithms for retrieving the cloud optical thickness (COT) and effective particle size (EPS) using the Visible/Infrared Imager/Radiometer Suite (VIIRS) channel brightness temperatures and reflectances. Input parameters required by the algorithms also include the VIIRS Cloud Mask, other VIIRS cloud Environmental Data Record (EDR) derived quantities (surface albedo, surface temperature, cloud top height), and scenario parameters (sun-sensor geometry, atmospheric sounding, surface type, spectral albedo/emissivity library, etc.). The overall processing begins with the identification of cloud-contaminated pixels and the determination of the cloud phase associated with the cloudy pixel via the VIIRS Cloud Mask/phase program. Four basic retrieval techniques are developed: solar and infrared (IR) for both cirrus and water clouds. The solar approach uses the reflectance of VIIRS 0.672, 1.2, 1.6, and 2.25 µm channels and follows the two-channel correlation technique (Nakajima and King, 1990) to retrieve the COT and EPS during the daytime. The IR approach utilizes the radiance of VIIRS 3.7 and 10.76 µm channels and follows the two-channel cirrus technique (Ou et al. 1993) to infer the cloud-top temperature and IR emissivity. Application of the IR approach during the local daytime requires the removal of the solar component in the 3.7 µm radiance. The COT and EPS can then be determined on the basis of the theory and parameterizations of radiative transfer and cloud microphysics. To carry out the algorithm sensitivity studies, we use a line-by-line equivalent (LBLE) radiative transfer model developed at UCLA to generate radiance tables for various combinations of COT and EPS. Algorithm sensitivity studies are composed of signal-to-noise ratio (SNR) tests as well as the error budget (EB) studies. The SNR tests use the UCLA-LBLE-generated radiance tables covering measurement range of both COT and EPS. Randomly distributed noises are added to these radiances according to the sensor noise model. These simulated radiances are then applied to retrieval algorithms. For EB studies, perturbation of the no-noise radiance tables for various error sources were carried out to simulate typical uncertainties in sensor parameters, surface characteristics, and atmospheric soundings. Using the Litton-TASC's Cloud Scene Simulation Model (CSSM) along with the perturbed radiance tables, cloud optical depth and associated reflectance/radiance fields are generated for several defined scenes. Error budgets are developed by applying these reflectance/radiance fields to retrieval algorithms. The methodologies and results of algorithm sensitivity studies on the retrieval of COT and EPS are documented herein, along with relevant practical considerations.

#### 1.0 INTRODUCTION

#### 1.1 PURPOSE

On May 5, 1994, President Clinton directed that the meteorological satellite programs operated separately by the National Oceanic and Atmospheric Administration (NOAA), U.S. Department of Commerce (DOC), and the Department of Defense (DoD) be converged into a single satellite program. Currently, the operating NOAA satellite system belongs to the Polar-orbiting Operational Environmental Satellite (POES) program, and the operating DoD satellite system belongs to the Defense Meteorological Satellite Program (DMSP). The future converged system, the National Polar-orbiting Operational Environmental Satellite System (NPOESS), will provide a single national remote sensing capability to acquire, receive, and disseminate global real-time weather data. The system will satisfy civil and national security operational requirements, and save more than \$1 billion in government spending.

The Visible/Infrared Imager/Radiometer Suite (VIIRS) is being developed as part of the NPOESS platform to meet the operational requirements for the global remote sensing of atmospheric and surface properties. One of the prime applications of the VIIRS channels is the remote sensing of cloud properties, including cloud cover/layers, cloud optical depth, cloud mean effective particle size, cloud top and base heights, cloud top temperature and pressure. These parameters are generally termed cloud environmental data records (EDR) in the NPOESS project.

This document is written to assist the sensor design and algorithm development particularly addressing the two NPOESS cloud EDRs: the cloud optical thickness (COT) and effective particle size (EPS). We present the theoretical basis for the solar and IR retrieval algorithms. We also describe the line-by-line equivalent (UCLA-LBLE) radiative transfer program that has been developed by our research group. LBLE was developed specifically for the purpose of simulating the clear and cloudy radiances for the algorithm development and sensitivity analyses using VIIRS channels

#### 1.2 SCOPE

This document covers the algorithm theoretical basis for the retrieval of the cloud optical thickness and effective particle size using VIIRS channel reflectances and radiances. Section 1 describes the purpose and scope of the document. Section 2 provides a scientific and historical background. The processing concept and algorithm description is presented in Section 3. Section 4 summarizes relevant assumptions and limitations. Finally, literature references used throughout the document are provided in Section 5.

#### 1.3 VIIRS DOCUMENTS

VIIRS Sensor Requirement Document (SRD)

VIIRS Error Budget Document (Y3249.v3)

#### 1.4 REVISIONS

This is the fourth (4<sup>th</sup>) revision of this document, dated May 2001.

#### 2.0 EXPERIMENT OVERVIEW

This section contains three major subsections. Subsection 2.1 describes the objectives of the cloud optical depth and effective particle size retrievals. Subsection 2.2 describes the characteristics of the VIIRS instrument. Subsection 2.3 addresses the cloud optical depth and effective particle size retrieval strategy.

## 2.1 OBJECTIVES OF CLOUD OPTICAL THICKNESS AND EFFECTIVE PARTICLE SIZE RETRIEVALS

Cloud optical thickness and effective particle size are two key NPOESS cloud EDRs. The objective of the development of cloud retrieval algorithms using the VIIRS cloud retrieval channels is the following: to achieve comprehensive retrieval of the cloud optical thickness and effective particle size accurately, meeting the threshold requirements and, in some cases, the objective requirements, as specified in the NPOESS Sensor Requirement Document (SRD).

The importance of clouds in weather and climate processes has been recognized as a result of numerous observational and modeling studies. Cloud microphysics, phase, and particle shape and size distributions determine the cloud optical properties. These optical properties affect the emission, transmission, reflection, and absorption of radiation by the cloud, and therefore modulate the balance of the atmospheric radiation budget. Clouds can be divided according to their particle phase into cirrus (ice) clouds and water (low) clouds. Both types of clouds are global in nature. Cirrus clouds occur primarily in the upper troposphere and the lower stratosphere, and are composed almost entirely of irregularly shaped, nonspherical ice crystals. Their global coverage is more than 30%. On the other hand, water clouds occur primarily in the middle and lower regions of the troposphere, and are composed of spherical water droplets. Their global coverage is more than 40%. Information on cirrus and water cloud parameters is critically important to the development of cloud forecast models, the upgrading of real-time global cloud analyses, the investigation of cloud feedbacks in global climate change, and the development of remote sensing of aerosols and surface parameters.

There are several well-documented and well-tested approaches to the satellite remote sensing of cloud optical thicknesses and effective particle sizes (e.g. King et al. 1996a, b; Minnis et al. 1998; Baum et al. 2000). All of these methods are based on the assumption of radiative transfer in plane-parallel clouds. These techniques exploit the spectral dependence of water and ice extinction of atmospheric radiation, using wavelengths at which absorption by water vapor and other gases is minimal, and at which the scattering and absorption by cloud particles are sensitive to the cloud particle size distribution. In the past, cirrus cloud parameters (optical depth, temperature, emissivity, etc.) have been inferred by using IR imaging and sounding channels (e.g., Szejwach, 1982; Arking and Child, 1985; Liou et al., 1990). In particular, Ou et al. (1993) developed a physical retrieval scheme using radiance data from AVHRR 3.7 µm and 10.9 µm channels to infer nighttime cirrus cloud parameters, including cloud temperature, optical depth, and mean effective ice crystal size, based on the theory of radiative transfer and parameterizations. This retrieval scheme has been applied to the nighttime AVHRR data collected during the First International Satellite Cloud Climatology Project Regional Experiment —Intensive Field Observation (FIRE-I IFO). For application of this IR retrieval scheme to daytime conditions, a numerical scheme for removing the solar component in the 3.7 µm radiance has been developed (Rao et al., 1995). The resulting removal-retrieval program has also been applied to AVHRR data collected during both FIRE-I and FIRE-II IFO (Ou et al., 1995).

Validation of the cirrus cloud parameters from this program has been carried out using co-located *in situ* ice crystal size distribution from 2-D probe measurements and ground-based lidar return imageries. Ou *et al.* (1996) developed a multiple threshold test scheme to detect cloud phase and multilayer cirrus cloud systems using AVHRR data, based on the physical properties of the AVHRR channel radiances. This scheme has been applied to the AVHRR data collected over the FIRE-II IFO area during nine overpasses within seven observational dates. Results from the cloud typing program have been verified using co-located and coincident ground-based radar and lidar return images (e.g., Clothiaux *et al.*, 1995), balloon-borne replicator data, and National Center for Atmospheric Research (NCAR) Cross-chain LORAN Atmospheric Sounding System (CLASS) humidity soundings on a case-by-case basis. Subsequently, the daytime IR retrieval program was modified for the retrieval of cirrus cloud parameters in multi-layer cloud systems. Validation of the multilayer cirrus cloud retrieval has been performed for two dates during FIRE-II IFO.

A number of studies on the determination of water cloud optical thicknesses and effective droplet radius during daytime have also been carried out using radiances from airborne visible and near-IR radiometers on aircraft (e.g., Hansen and Pollack, 1970; Twomey and Cocks, 1982 and 1989; Nakajima and King, 1990). The underlying principle on which these techniques are based is the fact that the reflection function of clouds at a non-absorbing channel in the visible wavelength region is primarily a function of the cloud optical thickness, whereas the reflection function at a water (or ice) absorbing channel in the near-infrared (e.g. 0.16 and 2.13µm channels) is primarily a function of cloud particle size (King et al., 1996). These studies have demonstrated the applicability of remote sensing methods to the determination of cloud optical and microphysical properties for water clouds using visible and near-IR channel reflectances. The principle of this two-channel correlation technique is now being applied to the development of an algorithm for the determination of cirrus microphysical and optical properties using the MAS 0.657 and 1.609 µm channel reflectances (e.g., Rolland et al., 2000). Though initial success has been achieved for the retrieval of cloud optical and microphysical parameters using both IR and solar approaches, further algorithm sensitivity and validation studies are required in order to assess the accuracy and precision of these methods when applied to measurements on a global scale.

The cloud optical depth and effective particle size retrieval algorithms, together with the prospective VIIRS sensor, are being developed to meet SRD requirements for the respective cloud EDRs. Under the VIIRS sensor/algorithm development concept, these requirements are "flowed down" to the design of the most cost-effective sensor/algorithm solution that meets the SRD requirements. This is accomplished through a series of flow down tests and error budget analyses, which effectively simulate sensor and algorithm performance over a range of environmental and operational scenarios. The error budgets are briefly described in Sections 3.4 and 3.5 and are described in much more details in the Raytheon VIIRS Error Budget, Version 3 (Y3249).

#### 2.1.1 Cloud Optical Depth

The SRD provides the following definition for Cloud Optical Depth:

Cloud optical depth is defined as the extinction (scattering + absorption) vertical optical thickness of all cloud layers in a vertical column of the atmosphere. Optical thickness ( $\tau$ ) is related to transmittance (t) by  $t = \exp(-\tau)$ . Optical thickness is wavelength dependent and is to be measured in at least two narrow bands centered at 450 nm (TBR) and 850 nm (TBR), with TBD nm bandwidth.

In the present retrieval algorithms, the retrieved cloud optical depth is defined as that at  $0.55 \mu m$ . Table A.1 in Appendix A summarizes the SRD requirements for this parameter.

#### 2.1.2 Cloud Effective Particle Size

The SRD provides the following definition for Cloud Effective Particle Size:

Effective cloud particle size is defined as the ratio of the third moment of the drop size distribution to the second moment, averaged over a layer of air within a cloud.

Table A.2 in Appendix A summarizes the SRD requirements for this parameter.

#### 2.2 INSTRUMENT CHARACTERISTICS

VIIRS is currently being designed based on cost, NPOESS requirements of the sensor, and the specified threshold/objectives requirements. Therefore, the following specifications of VIIRS are nominal and tentative. Table 1 lists the baseline bands and bandwidths for the cloud EDRs. Bands addressed in the cloud EDR algorithm sensitivity tests contained in this and other reports are marked with "x". Figure 1 shows the wavelength locations of seven VIIRS visible and near-IR channels used for retrievals of cloud effective particle size and optical thickness. The atmospheric zenith transmittances from altitudes 10, 5 and 0 km to the top of atmosphere are calculated from the UCLA-LBLE model based on the U.S. Standard Atmosphere. The solar zenith angle is 32°. Note that all these channels, except the channel at 1.378 μm, are located at window wavelengths associated with very high atmospheric transmittance. During the period of the project, slightly shifted central wavelengths and band widths have been used for algorithm development and sensitivity studies. Mainly for Bands I1, M11, and I4, we set central wavelengths at 0.645, 2.13 and 3.75 μm, respectively. Because the differences for cloud and gaseous optical properties due to these band shifts are small, we expect the results of algorithm studies using the shifted bands can be applicable to the bands specified in Table 1. Preliminary studies have been performed to verify this point.

At present, VIIRS is designed to scan through nadir in a plane perpendicular to the velocity vector of the spacecraft, with the maximum scan extending up to  $55^{\circ}$  on either side with respect to the spacecraft nadir. At a nominal orbital altitude for the NPOESS platform of about 833 km, this yields a swath width of 3000 km centered on the satellite nadir ground track. VIIRS is nominally composed of 12 channels. Three of the solar channels (0.672, 1.6, and 2.25  $\mu$ m) will be used in our solar cloud retrieval algorithm. One near-IR channel (3.7  $\mu$ m) and one thermal IR channel (10.8  $\mu$ m) will be used in the IR retrieval algorithm. Both the solar and IR channels are located in the water vapor window regions.

Table 1. VIIRS Baseline Bands and Bandwidths for the Cloud EDRs

VIIRS BAND	Lower Limit (µm)	Upper Limit (μm)	λ (μm)	Cloud Top Parameters	Cloud Effective Particle Size	Cloud Optical Depth	Cloud Mask/ Cover/ Layers
M1	0.401	0.419	0.410				
M2	0.436	0.454	0.445				
M3	0.478	0.498	0.488				
M4	0.545	0.565	0.555				
11	0.620	0.670	0.645				
M5	0.662	0.682	0.672		х	Х	х
M6	0.745	0.757	0.751				
12	0.845	0.885	0.865				
M7	0.845	0.885	0.865				х
M8	1.230	1.250	1.240		х	Х	
M9	1.371	1.386	1.378				х
13	1.580	1.640	1.610				
M10	1.580	1.640	1.610		x	Х	х
M11	2.225	2.275	2.250		x	Х	х
14	3.610	3.790	3.700				
M12	3.610	3.790	3.700	х	х	х	х
M13	3.973	4.128	4.050				
M14	8.400	8.700	8.550				х
M15	10.713	10.813	10.763	Х	x	Х	х
15	10.500	12.400	11.450				
M16	11.538	12.488	12.013				х

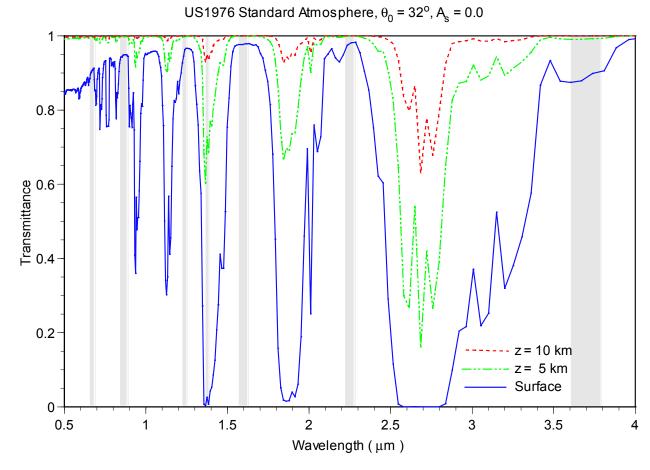


Figure 1. Spectral characteristics of the seven VIIRS solar channels, centered at 0.672, 0.859, 1.24, 1.378, 1.6, 2.25, and 3.7  $\mu$ m, used for retrievals of optical thickness and effective particle size. The atmospheric zenith transmittances are calculated from the LBLE model for three levels at 10 km, 5 km and surface based on the US Standard Atmosphere at 32° solar zenith angle.

#### 2.3 RETRIEVAL STRATEGY

A high-level flow diagram of the general approach to determining the COT and EPS parameters is provided in Figure 2. Input parameters required by these algorithms include the VIIRS Cloud Mask, other VIIRS cloud EDR derived quantities (e.g., surface temperature), VIIRS radiances, and scenario parameters (sun/sensor geometry, atmospheric scenario, etc.). The overall processing begins with the detection of cloud-contaminated pixels and the determination of their associated phase via the VIIRS cloud mask program. For each cloud phase, retrieval algorithms contain two basic retrieval approaches: solar and infrared (IR). The solar approach uses the reflectance of the 0.672, 1.6 and 2.25 µm channels for daytime retrieval of COT and EPS. It determines these cloud parameters by matching measured reflectances with those from the comprehensive radiance look-up tables, which are to be constructed using the pre-computed radiances from the LBLE for a wide range of scenarios. The IR approach utilizes the radiance of VIIRS 3.7 and 10.8 µm channels to infer cloud temperature and IR emissivity, from which the COT and EPS can be determined on the basis of the theory of radiative transfer and parameterizations.

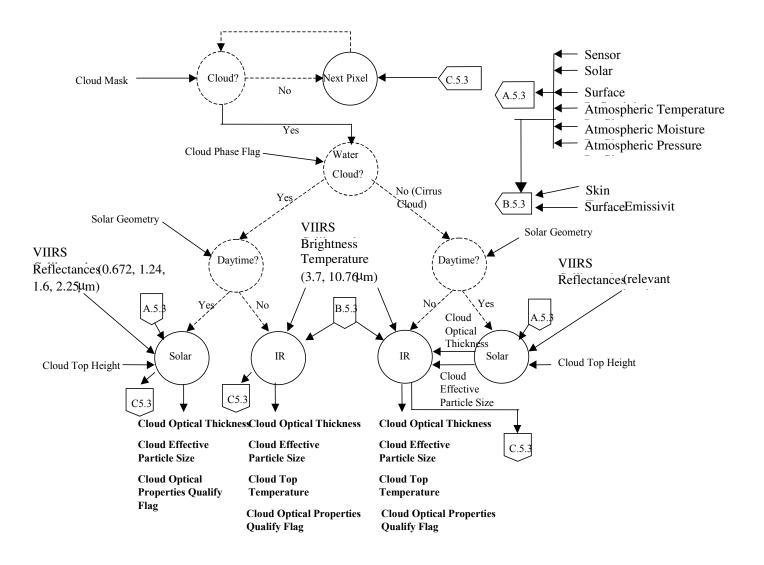


Figure 2. A high-level flow chart describing the general retrieval strategy for cloud optical thickness and effective particle size.

#### 3.0 ALGORITHM DESCRIPTION

#### 3.1 PROCESSING OUTLINE

The development of algorithms for the retrieval of cloud optical thickness (or optical depth) and effective particle size using the VIIRS channel reflectances/radiances involves a series of processing steps. As shown in Figure 3, the first step in the process involves defining scenarios to be used in algorithm sensitivity tests. These scenarios form the basis for defining the input parameters required by the radiative transfer model. The input parameters include climatic (or sounding) atmospheric profiles, sun-sensor geometries, cloud phases, surface properties, gaseous absorption properties, and cloud radiative properties. In the algorithm development and sensitivity studies for the retrieval of cirrus and water cloud parameters, the LBLE radiative transfer model is used, which is described in section 3.3.1.7. The LBLE is the most advanced code for computing high spectral resolution radiance using the combination of the correlated-*k* distribution (CKD) method and the adding-doubling method. The results from the LBLE have been compared with those from the MODTRAN, as described in section 3.3.1.7. It is shown that for larger cirrus cloud optical depths, the radiance from LBLE differs greatly from that from MODTRAN, mainly because MODTRAN assumes that the radiance is azimuthally independent.

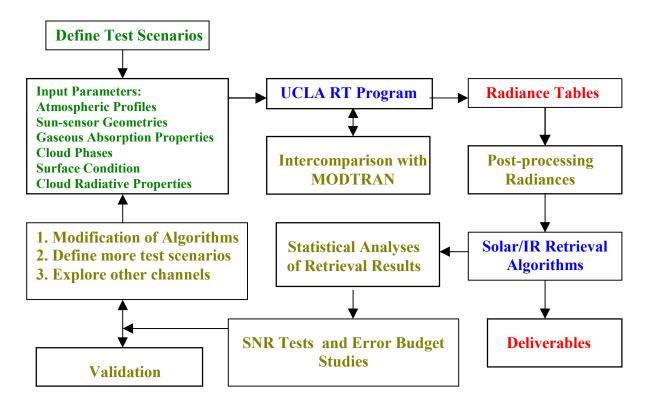


Figure 3. A high-level flow diagram depicting the processing outline of the data flow of the VIIRS channel radiance in the development of VIIRS cloud EDR retrieval algorithms.

A given specification for a scenario includes an atmospheric profile, solar and sensor geometry, band parameters, and cloud phase and altitude. For each scenario, the LBLE is executed over a wide range of optical thicknesses and effective particle sizes. Results for each scenario and for each spectral band are stored in the format of radiance tables. These radiance tables are used throughout the algorithm sensitivity studies. Post-processing procedures have been developed to convert the numbers in radiance tables into appropriate forms for the cloud parameter retrieval algorithms. For the solar algorithm, the radiance values are converted into reflectance using the solar spectral irradiance at the top of atmosphere. For the IR algorithm, the radiances are converted into spectrally averaged intensities and brightness temperatures. In the algorithm sensitivity studies, radiances are further processed to simulate instrument noises or calibration errors. For example, in the signal-to-noise (SNR) sensitivity analyses, noises are added to radiances according to the sensor noise model. Synthetic retrievals are carried out using the set of converted solar reflectances or IR intensities. The results of the retrieval processes are then statistically analyzed to compute the metrics described in the SRD (accuracy and precision errors, etc.). The metrics are then compared with the threshold/objective requirements specified in the SRD to determine if the retrieval algorithms meet these requirements.

If some part of the metrics fails the threshold/objective test, we perform an iterative process. There are three options: modify the algorithm to improve the performance, define more scenarios to search for the exact limit of application of the algorithms, or use additional channel data to improve the algorithm behavior. Finally, we validate retrieval algorithms by applying airborne (e.g. MAS) or satellite (e.g. AVHRR and MODIS) imagery data to the algorithms. MAS, AVHRR, and MODIS data have similar channel characteristics to those of VIIRS. We will compare retrieval results using imagery data collected during various field campaigns with collocated *in situ* microphysical and ground-based radiation measurements.

#### 3.2 ALGORITHM INPUT

The required input parameters for the retrieval of cloud optical thickness and effective particle size come from VIIRS and non-VIIRS sources and are summarized individually in the following paragraphs. The VIIRS data either come directly from the VIIRS processing stream, or are generated from the radiative transfer model.

#### 3.2.1 VIIRS Data

#### 3.2.1.1 Calibrated Brightness Temperatures and Reflectances

These data are required for input into both solar and IR algorithms.

#### 3.2.1.2 Geometric Parameters

These include solar and sensor-viewing zenith and azimuthal angles with respect to the target normal. Azimuthal angles are referenced to the local east. These inputs are obtained from simulated VIIRS orbital information. Information of solar zenith angle determines whether daytime or nighttime algorithms will be used.

#### 3.2.1.3 Exo-Atmospheric Solar Spectral Irradiances

These are required to convert radiance to bi-directional reflectance factors, which serve as input to the solar algorithm. The solar constant for each VIIRS cloud retrieval band is determined based on the solar irradiance spectrum used in the MODTRAN v3.7.

#### 3.2.1.4 Cloud Mask Product

The identification of cloudy/clear pixels and associated cloud phases may be obtained from the VIIRS cloud mask product. Cloud/clear information will be used to filter out clear pixels.

#### 3.2.1.5 VIIRS Band Parameters

These are required for the derivation of single-scattering properties of cloud particles, and for the construction of radiance look-up tables.

#### 3.2.1.6 Cloud Top Height

These are required for the construction of radiance look-up tables for solar algorithms.

#### 3.2.1.7 Surface Type

This information will help the determination of surface albedo for the solar algorithm.

#### 3.2.1.8 Surface Temperature

This information is needed for the computation of VIIRS IR channel radiances.

#### 3.2.2 Non-VIIRS Data

#### 3.2.2.1 Atmospheric Sounding

In the construction of radiance look-up tables and in the execution of retrieval algorithms, it is necessary to define the atmosphere by using the atmospheric sounding to compute the clear and cloudy radiances. Atmospheric sounding includes temperature, pressure and moisture profiles.

#### 3.2.2.2 Spectral Library

The spectral library contains spectral reflection and emission properties for various surface types and is used in conjunction with the VIIRS surface type information to determine surface reflectivity and emissivity.

## 3.3 THEORETICAL DESCRIPTION OF OPTICAL DEPTH AND EFFECTIVE PARTICLE SIZE RETRIEVALS

In section 3.3.1, we will discuss the physics of the problem, including the characteristics of cloud particle size distribution, radiative properties of clouds, parameterization of atmospheric gaseous absorption, modeling of radiative transfer within clouds, and the inter-comparison of computed radiances from the LBLE and the MODTRAN for both clear and cloudy atmospheres. Section 3.3.2 focuses on the mathematical description of retrieval algorithms.

#### 3.3.1 Physics of the Problem

Developing accurate and reliable cloud remote sensing programs requires that a detailed and accurate radiative transfer program be available for algorithm development and sensitivity studies. To simulate the atmospheric scattering/absorption effects accurately, it is essential that reliable phase functions and single-scattering properties for non-spherical ice crystals, spherical water droplets, molecules and aerosols be used. Single-scattering properties include the extinction coefficient (or efficiency), the single-scattering albedo, and the asymmetry factor. It is also important to incorporate an efficient and accurate parameterization program for computing atmospheric gaseous absorption/emission coefficients. Section 3.3.1.1 presents the derivation of typical optical properties of cirrus clouds. We establish representative ice crystal size distributions by analyzing the observed ice crystal size distributions that have become available from field experiments in the mid-latitude region (e.g. FIRE-I IFO). The objective is to characterize the ice crystal size distribution in terms of sizes and shapes that are representative of the mid-latitude cirrus for remote sensing applications. Section 3.3.1.1 discusses the calculation of the scattering and absorption properties of ice crystals of various sizes and shapes based on the geometric-optics/integral-equation and finite-difference time domain methods. Sections 3.3.1.2, 3.3.1.3, and 3.3.1.4 address the determination of optical properties of water droplets, molecules and aerosols, respectively. Section 3.3.1.5 describes the parameterization of atmospheric gaseous absorption by way of the correlated-k distribution approach. Finally, section 3.3.1.6 explains how the single-scattering properties for combined cloud particles and gases are derived.

The LBLE radiative transfer model is employed to carry out the algorithm development and sensitivity studies. Section 3.3.1.7 describes the physical principles and model structure of the LBLE. Results of inter-comparison between LBLE and MODTRAN-DISORT (MD) are presented in this section as well. It is shown that, in general, the MD radiance for optically thick clouds differs from the LBLE radiance by more than 10 percent. This difference increases with increasing optical depths. The present retrieval program is closely coupled with the LBLE. If the future weather product test bed uses the MD or a program with similar philosophy for modeling multiple scattering, it is expected that the retrieved cloud EDRs would not meet the EDR threshold/objective requirements.

#### 3.3.1.1 Microphysical and Radiative Properties of Cirrus Clouds

#### Analysis of the observed ice crystal size distributions

Development of the remote sensing algorithms for ubiquitous cirrus clouds requires comprehensive knowledge of two fundamental ice microphysics parameters: ice crystal size distribution and ice crystal shape. Both vary in space and time, and are associated with microphysical processes, including diffusional growth/shrinkage by water vapor deposition/sublimation, and aggregation by collision and coalescence between ice crystals. It is necessary to define representative ice crystal size

distributions and shapes for the calculations of phase functions and other relevant single-scattering properties used in satellite remote sensing applications. In conjunction with our ongoing radiative transfer and satellite remote sensing studies involving cirrus clouds, we have analyzed six composite ice crystal size distributions that were obtained from aircraft observations presented by Heymsfield and Platt (1984), Takano and Liou (1989), and the FIRE-IFO microphysical data. They are denoted as cold Ci, -60° C, Cs, FIRE-I IFO 1 Nov, FIRE-I IFO 2 Nov, and Ci uncinus. Figure 4 shows the six ice crystal size distributions. The ice crystal sizes span from about 5 to 2000 µm with shapes ranging from bullet rosettes, solid and hollow columns, and plates to aggregates. We define a mean effective size to characterize ice crystal size distribution in the form:

$$D_e = \int D^2 L \, n(L) \, dL / \int D \, L \, n(L) \, dL \tag{1}$$

where D and L denote the width and the maximum dimension of a non-spherical ice crystal, respectively, and n(L) is the size distribution in terms of L. The rationale for defining  $D_e$  to represent ice crystal size distribution is that the scattering of light is related to the geometric cross section, which is proportional to LD. The preceding definition of the mean effective size is applicable to irregular and complex ice crystal shapes (Ou et al., 1995). In the present algorithm development and sensitivity studies, effective particle size for ice crystals is defined as half of the mean effective size. This definition is consistent with that in the SRD, because  $D_e$  is equivalent to a measure of diameter, while the effective particle size is equivalent to radius. At this point, sufficient information concerning ice crystal habit and percentage of individual habit for a given size distribution is not available. We propose ice crystal habit and associated aspect ratios on the basis of the in situ 2-D probe and replicator data. Following Takano and Liou (1989a), we have aggregated the size distributions into 5 bins: 10-30 µm with bin-center at 20 µm, 30-70 µm with bin center at 50 µm, 70 – 170  $\mu$ m with bin-center at 120  $\mu$ m, 170-430  $\mu$ m with bin-center at 300  $\mu$ m, and 430 – 1070  $\mu$ m with bin center at 750  $\mu$ m. The aspect ratios, L/D, used are 20/20, 50/40, 120/60, 300/100, and 750/160 (in units µm/µm). These roughly correspond to the observations reported by Auer and Veal (1970). We have evaluated the mean effective ice crystal size for the six size distributions, which are presented in Table 2.

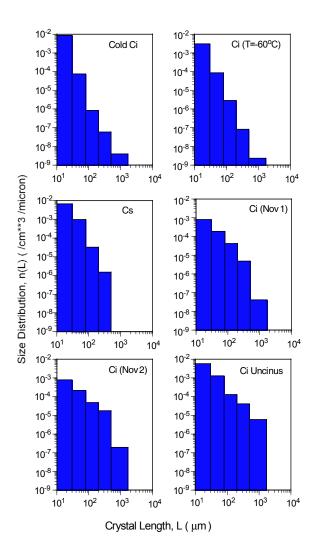


Figure 4. The six discrete ice crystal size distributions used in generating radiances and algorithm sensitivity studies.

Table 2. Mean Effective Size of the Six Ice Crystal Size Distributions Employed in the Present Work

Ice crystal size distribution	Mean effective size (μm)		
Cold Ci	23.9		
T = -60 ° C	30.4		
Cs	41.5		
FIRE-I Nov. 1	75.1		
FIRE-I Nov. 2	93.0		
Ci Uncinus	123.6		

#### Single-scattering properties for ice crystals

Unlike the scattering of light by spherical water droplets, which can be solved by the exact Mie theory, an exact solution for the scattering of light by hexagonal ice crystals covering all shapes (habits) and sizes does not exist in practical terms. Although several numerical methods have been developed to solve the problem of light scattering by nonspherical particles, they are usually applicable to size parameters (ka, where k is the wave number and a is the semi-width of an ice crystal) smaller than approximately 20, as discussed by Liou and Takano (1994), Yang and Liou (1995), and the references cited therein. In the past two decades, significant research on solving light scattering by regular and complex ice crystals has been carried out by means of the geometric ray-tracing technique, commonly employed to identify the optical phenomena occurring in the atmosphere. In the limit of geometric optics, an incident wave may be considered as being composed of a bundle of rays that strike the ice crystal and undergo reflection and refraction along a straight line. The laws of geometric optics are applicable to the scattering of light by an ice crystal if its size is much larger than the incident wavelength. In those cases the geometric rays can be localized.

In recent years, we have developed a Monte Carlo/geometric ray-tracing method for the computation of the scattering, absorption, and polarization properties of ice crystals with various regular and irregular structures. These structures include solid and hollow columns, single and double plates, dendrites, bullet rosettes, and aggregates (Takano and Liou, 1995). The shapes of these ice crystals are defined by appropriate geometric models and incident coordinate systems. The incident photons are traced with a hit-and-miss Monte Carlo method and followed by geometric reflection and refraction on the crystal boundary. Absorption can be accounted for by means of stochastic procedures. For some of the quasi-spherical ice particles, such as frozen droplets, we can approximate them by spheroids. In those cases, the scattering and absorption properties can be computed exactly (Takano *et al.*, 1992).

In the geometric optic method, the extinction efficiency is set at 2. Thus the extinction coefficient is equal to twice the total cross sectional area per unit volume, and is spectrally independent for a given ice crystal size distribution. The single-scattering albedo is parameterized in terms of the absorption coefficient  $k_i=4\pi m_i/\lambda$  and the aspect ratio L/D:

$$\overline{\omega} = Q_s / Q_e = \sum_{n=0}^4 b_n z^n$$
 where  $z = k_i a \frac{3\sqrt{3}(L/D)}{\sqrt{3} + 4(L/D)}$  (2)

The scattering phase function is obtained from the phase matrix, which is determined using the Monte Carlo / geometric ray-tracing method. We assume that ice crystals are randomly oriented.

For non-spherical ice crystals, the phase function contains a strong forward peak at  $0^{\circ}$  scattering angle. The forward scattered energy within  $\sim 5^{\circ}$  scattering angles produced by diffraction is four to five orders of magnitude greater than it is in the side and backscattered directions. To incorporate the forward peak contribution in multiple scattering, we consider an adjusted absorption and scattering atmosphere, such that the fraction of scattered energy residing in the forward peak, is removed from the scattering parameters. These parameters include the optical depth, single-scattering albedo, and asymmetry factor. Thus the phase function is expressed as

$$f_D = \frac{1}{2\varpi (1 - f_{\delta})}, \text{ where}$$

$$P(\Theta) = (1 - f_D)P_G(\Theta) + f_D P_D(\Theta),$$
(3)

and  $f_D$  is the ratio of the diffracted light to the entire scattered light.

The geometric ray-tracing requires the principle of localization in which the particle size must be larger than the incident wavelength. In addition, it is assumed that the energy attenuated by the scatterer may be decomposed into equal extinction from diffraction and Fresnelian rays so that the extinction efficiency is equal to 2 regardless the particle size parameter. To circumvent a number of shortcomings in the geometric-optics approach, we have developed a novel improvement by mapping the equivalent tangential electric and magnetic currents on the particle surface, obtained from geometric reflection and refraction, to far-field by means of the basic electromagnetic wave theory in the two-dimensional (2-D) space (Yang and Liou, 1995). We have further extended the improved geometric-optics method in the 2-D case to three-dimensional (3-D) space, allowing arbitrary and random orientations of the ice crystals (Yang and Liou, 1996a). The improved geometric ray-tracing method is referred to as the geometric-optics/integral-equation hybrid method (GOM2).

For verification of the limitations of the geometric optics, we have also developed a finite-difference time domain (FDTD) technique, pioneered by electrical engineers for the identification of irregular objects. It is a numerical technique for the solution of the Maxwell equations using appropriate absorbing boundary conditions. It is considered to be the "exact" numerical solution for light scattering by particles, as verified by the exact Mie results for long circular cylinders and spheres (Yang and Liou, 1996b). Because of numerical instability and required computer time, the FDTD method can only be applied to size parameters smaller than about 20. We show that the GOM2 method converges to the conventional ray-tracing method for size parameters larger than about 200. The conventional method breaks down when the size parameters of ice crystals are smaller than about 40-100, depending on whether the computations are for the cross sections or phase matrix. Moreover, based on comparison with the results computed from the FDTD method, the GOM2 technique is shown to be applicable to ice crystal size parameters as small as 15-20.

As pointed out previously, an exact solution for the scattering of light by hexagonal ice crystals covering all shapes and sizes does not exist in practical terms. However, by combining the GOM2 (ka > 20) and FDTD (ka < 20) methods, we have developed a unified theory for light scattering by ice crystals essentially for all sizes and shapes.

At this point, we have the theoretical tools to provide the basic scattering and absorption data required for a reliable simulation of the single-scattering properties of ice clouds. The unified theory for light scattering, along with representative ice crystal size distributions and shapes discussed in Section 3.3.1.1 are used to compute the phase function and single-scattering properties for all VIIRS cloud retrieval wavelengths listed in Table 1. The computations are quite extensive and require a careful selection of sizes and shapes. The influence of the mean effective size and variance on the single-scattering properties can both be examined with respect to the potential remote sensing of these parameters. Shown in Figure 5 are the phase functions for the six size distributions and for the five VIIRS cloud retrieval channels obtained from the Monte Carlo/Geometric ray tracing method. For the solar channels, the overall feature of phase function is not sensitive to the variation of size distribution

because of the negligible absorption involved. The  $22^{\circ}$  and the  $46^{\circ}$  halos produced by two refracted rays are well illustrated for the phase functions for 0.645, 1.6 and 2.13 µm channels, in addition to the forward diffraction peak. The magnitude of the forward scattering associated with diffraction varies with size distribution due to strong absorption. For scattering angles between about  $150^{\circ}$  and  $160^{\circ}$ , there is another peak for all sizes produced by rays undergoing double internal reflections. The magnitude of the side-scattering is larger for smaller ice crystals. For the thermal channel, the halos and the backscattering peaks all disappear due to strong absorption effects. Figures 6(a) and 6(b) show the single-scattering albedos and asymmetry factors associated with the six size distributions and the five VIIRS cloud retrieval channels. In general, the single-scattering albedo decreases with increasing wavelengths, while the asymmetry factor increases with increasing wavelengths. For the solar channels, the single-scattering albedo decreases with increasing  $D_e$ , and is larger than 0.5. For the 10.8 µm channel, the single scattering albedo varies between 0.4 and 0.5 due to particle absorption effects, and increases with increasing  $D_e$ . The asymmetry factor increases with increasing  $D_e$ , implying that as  $D_e$  increases, more fraction is scattered in the forward direction.

#### 3.3.1.2 Water Cloud Optical Properties

Water cloud properties for VIIRS solar channels have been generated using the Mie scattering code for water droplets. For a single water droplet, the extinction and scattering efficiencies are obtained as

$$Q_s = \frac{2}{x^2} \sum_{n=1}^{\infty} (2n+1)(|a_n|^2 + |b_n|^2), \tag{4}$$

$$Q_{e} = \frac{2}{x^{2}} \sum_{n=1}^{\infty} (2n+1) \operatorname{Re}(a_{n} + b_{n}), \tag{5}$$

where  $x = 2\pi r/\lambda$  (the size parameter), and  $a_n$  and  $b_n$  are coefficients in the expansion for scattered waves. The phase function for a single water droplet is obtained from:

$$P(\Theta) = 4\pi [i_1(\Theta) + i_2(\Theta)] / 2k^2 \sigma_s \tag{6}$$

where  $i_1$  and  $i_2$  are the square of the magnitude of the scattering functions, k is the wave number  $(2\pi/\lambda)$ , and  $\sigma_s$  is the scattering cross section.

A gamma distribution is assumed for the droplet size distribution. Subsequently, the average extinction and scattering coefficients are obtained from:

$$\beta_{e,s} = \int_{r_1}^{r_2} Q_{e,s} \pi r^2 n(r) dr , \qquad (7)$$

where n(r) is the droplet size distribution in terms of radius r, and  $r_1$  and  $r_2$  are the minimum and maximum radius in the size distribution. The average single scattering albedo is then obtained as:

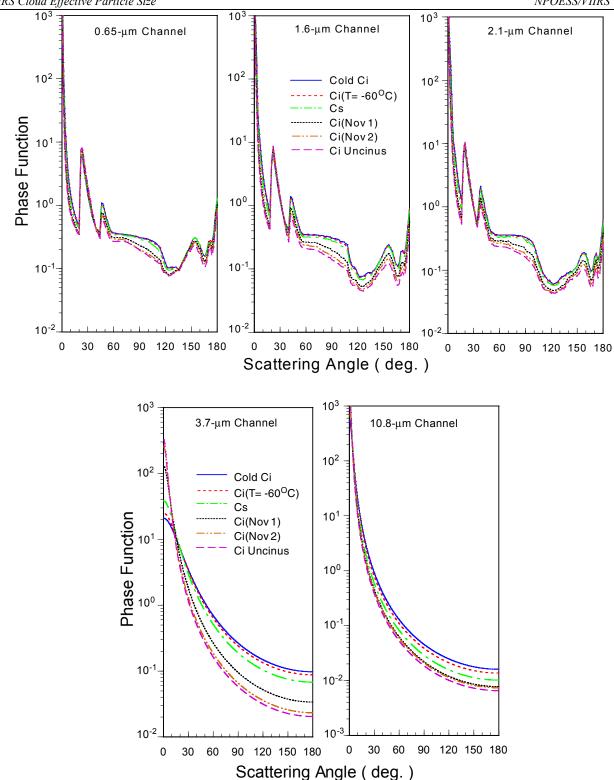


Figure 5. Phase functions for the six size distributions and for the five VIIRS cloud retrieval channels

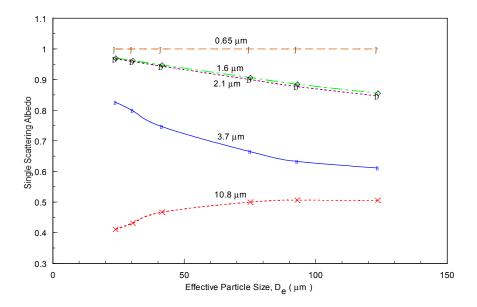


Figure 6a. Single scattering albedos for the six size distributions and for the five VIIRS cloud retrieval channels.

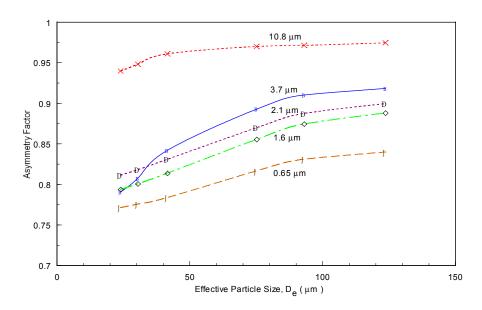


Figure 6b. Asymmetry factors for the six size distributions and for the five VIIRS cloud retrieval channels.

$$\overline{\varpi} = \frac{\beta_s}{\beta_e}.$$
 (8)

Finally, the average phase function is obtained from

$$P(\Theta) = \frac{4\pi}{k^2 \beta_s} \int_{r}^{r^2} \frac{1}{2} [i_1(\Theta, r) + i_2(\Theta), r)] n(r) dr, \qquad (9)$$

For VIIRS IR channels, we use an analytic approximation developed by Chylek *et al.* (1995), which is accurate to within a few percent when compared to exact Mie computations. Chylek's formulation is appropriate for wavelengths greater than 3 µm. Compared to Platnick's data (King *et al.*, 1996) at certain overlapping wavelengths, Chylek's formulation seems to be in good agreement.

#### 3.3.1.3 Optical Properties for Rayleigh Scattering

For very small size parameters (<< 1), the effect of particle size is small, and the scattered intensity is inversely proportional to the fourth power of wavelength. This scattering regime is referred to as the Rayleigh scattering. The scattering of solar radiation by air molecules falls within the Rayleigh scattering regime, which must be accounted for in the calculation of visible radiance.

In the Rayleigh scattering regime, the single–scattering albedo is set to be 1 (conservative scattering). Following the classical solution by Rayleigh (1918), the phase function for incident unpolarized solar radiation is:

$$P(\Theta) = \frac{3}{4}(1 + \cos^2 \Theta). \tag{10}$$

The Rayleigh optical depth for the Earth's atmosphere is spectrally dependent, and can be computed according to the following parameterization (Hansen and Travis, 1974):

$$\tau = 0.008569\lambda^{-4}(1 + 0.0113\lambda^{-2} + 0.00013\lambda^{-4}), \tag{11}$$

where  $\lambda$  is the wavelength. The above equation is derived for the standard surface pressure  $P_0 = 1013.25$  mb. For  $\lambda = 0.55 \mu m$ , the optical depth is about 0.0973.

#### 3.3.1.4 Aerosol Scattering

The averaged extinction coefficient, single-scattering albedo, and asymmetry factor for individual aerosol particles are obtained in the same manner as those parameters for water droplets. However, we use the Henyey-Greenstein phase function (Liou, 1992) for approximation:

$$p_{HG}(\Theta) = \frac{(1-g^2)}{(1+g^2 - 2g\cos\Theta)^{\frac{3}{2}}},$$
(12)

where g is the asymmetry factor.

### 3.3.1.5 Correlated-k Distribution Approach

The gaseous absorption coefficient is a function of wave number, pressure, and temperature, and can be written in terms of line strength S(T) and line shape function f(v, p, T) in the form:

$$k(v, p, T) = \sum_{i} S_{i}(T) f_{i}(v, p, T)$$
(13)

For a given absorption gas and spectral interval,  $\Delta v$ , we introduce the k-distribution function h(k), which is the probability density function such that h(k)dk is the fraction of  $\Delta v$  within which the absorption coefficient is between k and k + dk. Thus the spectral-mean transmittance should depend on the k-distribution but is independent of the ordering of the absorption coefficients, k(v), with respect to the wave number (Ambartzumian, 1936; Arking and Grossman, 1972). We may replace the wave number integration by integrating over the k-space. Let the maximum and minimum absorption coefficient within  $\Delta v$  be  $k_{max}$ , and  $k_{min}$ . Setting  $k_{min} \to 0$  and  $k_{max} \to \infty$  for mathematical convenience, the spectral-mean transmittance as a function of path length, u, may be expressed by:

$$T_{\overline{\nu}}(u) = \int_{\Delta \nu} e^{-k(\nu)u} \frac{d\nu}{\Delta \nu} = \int_{0}^{\infty} e^{-ku} h(k) dk, \qquad (14)$$

where h(k) is normalized to 1 in the domain  $(0, \infty)$ . Equation 14 defines the k-distribution approach, which is exact for the case of homogeneous atmosphere. We may further define a cumulative probability function in the form:

$$g(k) = \int_{0}^{k} h(k)dk , \qquad (15)$$

where g(0) = 0,  $g(k \rightarrow \infty) = 1$ , and dg(k) = h(k)d(k). By definition, g(k) is a monotonically increasing and smooth function in k space. It follows that the spectral-mean transmittance can now be expressed in terms of this cumulative probability function g(k) in the form:

$$T_{\overline{v}}(u) = \int_{\Delta v} e^{-k(v)u} \frac{dv}{\Delta v} = \int_{\mathbb{R}} e^{-k(g)u} dg, \qquad (16)$$

where k(g) is referred to as the equivalent k function, which is the inverse function of g(k). Since g(k) is a smooth and monotonically increasing function in k space, k(g) must also be a smooth and monotonically increasing function in g space.

In order to apply the k-distribution method to a nonhomogeneous atmosphere, we shall consider this atmosphere defined by two heights,  $z_1$ , and  $z_2$ . The spectral-mean transmittance can be written in the form:

$$T_{\overline{v}} = \int_{\Delta v} \exp\left[-\int_{z_1}^{z_2} k(v, p, T) \rho dz\right] \frac{dv}{\Delta v}$$
(17)

where  $\rho$  is the density of the absorber and we have introduced pressure and temperature dependence in the absorption coefficient. We wish to investigate the physical and mathematical conditions under which Equation 17 may be expressed in the form similar to Equation 16, namely:

$$T_{\bar{v}} = \int \exp[-\int_{z_1}^{z_2} k(g, p, T) \rho dz] dg$$
 (18)

The method for calculating spectral-mean transmittance based on Equation 18 is referred to as the correlated k-distribution method (CKD). In essence, the physical implication of CKD is that only one g exists for a given v at different heights.

Fu and Liou (1992) compared the distribution of the absorption coefficient in the wave number domain and in the cumulative probability (g) domain for the  $O_3$  9.6  $\mu$ m band at p=25mb and T=220K. They show that the absorption coefficient is a high-frequency oscillatory function of wave number, but is smooth in the g-domain. Thus, the efficient integration in the g-domain replaces the cumbersome wave number integration, and the associated numerical integration can be carried out by evaluating the integral at only a few selected g values.

Evaluation of the spectral-mean transmittance in nonhomogeneous atmospheres requires coupled height and wave number integrations (Equation 17). Traditionally, these integrations have been separated by some approximations such that band models can be applied. In this sense, one-parameter scaling, two-parameter Curtis-Godson approximation (Liou, 1992), and three-parameter scaling are all based on the same philosophy. However, in the CKD approach, after the transformation of  $\nu$  domain to g domain, the separation of height and g integrations is not necessary because the height integration can be effectively carried out in g space under the correlated assumptions. The approach fundamentally differs from traditional band models and scaling approximations and appears to offer a direct solution to the intricate radiative transfer problem involving nongray gaseous absorption.

# 3.3.1.6 Single-Scattering Properties for Combined Cloud and Aerosol Particles, Rayleigh Molecules, and Gases

We divide a nonhomogeneous atmosphere into a number of layers, each of which is assumed to be homogeneous. For the case when attenuation by cloud particles, Rayleigh scattering, and aerosols are occurring along with gaseous absorption, the total optical depth for each layer is:

$$\Delta \tau(g) = \Delta \tau^C + \Delta \tau^R + \Delta \tau^A + \Delta \tau^G(g), \tag{19}$$

where  $\Delta \tau^C$ ,  $\Delta \tau^R$  and  $\Delta \tau^A$  represent the optical depth due to cloud particles, Rayleigh molecules, and aerosols, respectively;  $\Delta \tau^G(g)$  is the optical depth contributed by the gaseous absorption for a given g (the cumulative probability), which can be expressed by:

$$\Delta \tau^{G}(g) = k(g, p, T) \rho \Delta z \tag{20}$$

where k(g,p,T) is the equivalent k function (Fu and Liou, 1992),  $\rho$  is the density of the absorber, and  $\Delta z$  is the geometric thickness of the layer. In this case,  $\Delta \tau^C = \Delta \tau_s^C + \Delta \tau_a^C$ , where  $\Delta \tau_s^C$  and  $\Delta \tau_a^C$  are cloud scattering and absorption optical depths, respectively. In the same manner,  $\Delta \tau^A = \Delta \tau_s^A + \Delta \tau_a^A$ ,

where  $\Delta \tau_s^A$  and  $\Delta \tau_a^A$  are aerosol scattering and absorption optical depths, respectively. Thus the combined single-scattering albedo can be obtained by:

$$\varpi(g) = \frac{\Delta \tau_s^C + \Delta \tau^R + \Delta \tau_s^A}{\Delta \tau^C + \Delta \tau^R + \Delta \tau^A + \Delta \tau^G(g)}$$
(21)

Similarly, when the phase function is applied to the radiative transfer program, it is expanded in the form:

$$P(\Theta) = \sum_{l=0}^{N} \boldsymbol{\sigma}_{l} P_{l}(\cos \Theta)$$
 (22)

where  $P_l(\cos\Theta)$  is the Legendre polynomial of the order l, and  $\varpi_l$  is the expansion coefficient. The expansion coefficient for the composite phase function can be obtained by:

$$\boldsymbol{\sigma}_{l} = \frac{\Delta \tau_{s}^{C} \boldsymbol{\sigma}_{l}^{C} + \Delta \tau^{R} \boldsymbol{\sigma}_{l}^{R} + \Delta \tau_{s}^{A} \boldsymbol{\sigma}_{l}^{A}}{\Delta \tau_{s}^{C} + \Delta \tau^{R} + \Delta \tau_{s}^{A}}$$
(23)

where  $\varpi_l^C$ ,  $\varpi_l^R$ , and  $\varpi_l^A$  denote the expansion coefficients for the individual phase function for cloud particles, Rayleigh molecules, and aerosols, respectively. Since the phase function is independent of gaseous absorption, the combined  $\varpi_l$  is constant over a given spectral absorption band. Once the single-scattering properties have been defined for a given g for each level, monochromatic radiative transfer calculations may be carried out, and the radiance for each spectral subinterval ( $\Delta v = 50 \text{ cm}^{-1}$ ) can be obtained as the summation of the radiance component in the g-domain:

$$R = \sum_{i=1}^{30} R(g_i) \Delta g_i \tag{24}$$

### 3.3.1.7 Radiative Transfer Model

### Model structure and flow chart

The radiative transfer scheme used is the adding-doubling method including full Stokes parameters developed by Takano and Liou (1989) for vertically inhomogeneous atmospheres. In an anisotropic medium, the single-scattering properties depend on the direction of the incoming light beam. Let the directions of the incoming and outgoing light beams be denoted by  $(\mu', \phi')$  and  $(\mu, \phi)$ , respectively, where  $\mu$  is the cosine of the zenith angle and  $\phi$  the corresponding azimuthal angle. The scattering phase matrix P is a function of  $(\mu, \phi, \mu', \phi')$  and cannot be defined by the scattering angle  $\Theta$  alone as in the conventional radiative transfer. Moreover, the extinction and scattering cross sections vary with the direction of the incoming light beam  $(\mu', \phi')$ .

Let the Stokes vector intensity I=(I, Q, U, V). Following Liou (1992), the general equation governing the transfer of diffuse solar intensity may be written in the form:

$$\mu \frac{d\mathbf{I}(\tau; \mu, \phi)}{d\tau} = \mathbf{I}(\tau; \mu, \phi) - \mathbf{J}(\tau; \mu, \phi)$$
(25)

and the source function,

$$J(\tau;\mu,\phi) = \frac{1}{4\pi} \int_{0}^{2\pi} \int_{-1}^{2\pi} \mathbf{P}(\mu,\phi;\mu',\phi') J(\tau;\mu',\phi') d\mu' d\phi'$$

$$+ \frac{1}{4\pi} P(\mu,\phi;-\mu_{o},\phi_{o}) \times \pi F_{o} \exp[-k(-\mu_{o})\tau/\mu_{o}]$$
(26)

In Equation 26,  $\mu_0$  is the cosine of the solar zenith angle,  $\phi_0$  the corresponding azimuthal angle, and -  $\mu_0$  denotes the downward solar incident direction. The first and second terms on the right-hand side represent contributions from multiple scattering and single scattering of the direct solar intensity, respectively.

The multiple scattering problem is solved by means of the adding-doubling principle for radiative transfer. We define the reflection matrix  $\mathbf{R}(\mu, \phi, \mu', \phi')$  and transmission matrix  $\mathbf{T}(\mu, \phi, \mu', \phi')$  for radiation from above in the forms:

$$I_{out,top}(\mu,\phi) = \frac{1}{\pi} \int_{0}^{2\pi} \int_{0}^{1} \mathbf{R}(\mu,\phi;\mu'',\phi'') I_{in,top}(\mu'',\phi'') \mu'' d\mu'' d\phi''$$
(27)

$$I_{out,top}(\mu,\phi) = \frac{1}{\pi} \int_{0}^{2\pi} \int_{0}^{\pi} T(\mu,\phi;\mu'',\phi'') I_{in,top}(\mu'',\phi'') \mu'' d\mu'' d\phi''$$
(28)

Likewise, for radiation from below, the reflection and transmission matrices are defined by:

$$\boldsymbol{I}_{out,bottom}(\mu,\phi) = \frac{1}{\pi} \int_{0}^{2\pi} \int_{0}^{1} \boldsymbol{R} * (\mu,\phi;\mu'',\phi'') \boldsymbol{I}_{in,bottom}(\mu'',\phi'') \mu'' d\mu'' d\phi'$$
(29)

$$I_{out,top}(\mu,\phi) = \frac{1}{\pi} \int_{0}^{2\pi} \int_{0}^{1} T *(\mu,\phi;\mu'',\phi'') I_{in,bottom}(\mu'',\phi'') \mu'' d\mu'' d\phi''$$
(30)

To proceed with the adding principle for the radiative transfer in an anisotropic medium, we shall utilize the reflection and transmission matrices defined in Equations 27 through 30 and consider an infinitesimal layer with a very small optical depth  $\Delta \tau$ , say  $10^{-8}$ . Since the optical depth is so small, only single scattering takes place within the layer. From the fundamental equations for the radiative transfer, the analytic solutions for reflected and transmitted intensities undergoing single scattering may be derived. Subject to the condition that  $\Delta \tau \rightarrow 0$ , we find:

$$\mathbf{R}(\mu,\phi;\mu',\phi') \approx \frac{\Delta \tau}{4\mu\mu'} \boldsymbol{\varpi} \mathbf{P}(-\mu,\phi;\mu',\phi') \tag{31}$$

$$T(\mu, \phi'; \mu', \phi') \approx \frac{\Delta \tau}{4\mu\mu'} \boldsymbol{\varpi} \boldsymbol{P}(-\mu, \phi; -\mu', \phi')$$
(32)

$$\mathbf{R} * (\mu, \phi; \mu', \phi') \approx \frac{\Delta \tau}{4\mu\mu'} \boldsymbol{\varpi} \mathbf{P}(\mu, \phi; -\mu', \phi')$$
(33)

$$T * (\mu, \phi; \mu', \phi') \approx \frac{\Delta \tau}{4\mu\mu'} \varpi P(\mu, \phi; \mu', \phi')$$
(34)

Consider now two layers denoted by subscripts a and b, where layer a is on top of layer b. Let their optical depths be  $\tau_a$  and  $\tau_b$ . Following the conventional adding principle for radiative transfer in an isotropic medium, but with modifications to account for the dependence of the optical properties on the incoming direction, the procedure for computing the reflection and transmission matrices for the composite layer may be described by the following equations:

$$\mathbf{Q}_1 = \mathbf{R}_a^* \mathbf{R}_b \tag{35}$$

$$\mathbf{Q}_{n} = \mathbf{Q}_{1}\mathbf{Q}_{n-1} \tag{36}$$

$$S = \sum_{n=1}^{M} \mathbf{Q}_n \tag{37}$$

$$\mathbf{D} = \mathbf{T}_a + \mathbf{S} \exp[-\tau_a / \mu_0] + \mathbf{S} \mathbf{T}_a \tag{38}$$

$$\boldsymbol{U} = \boldsymbol{R}_b \exp[-\tau_a / \mu_a] + \boldsymbol{R}_b \boldsymbol{D} \tag{39}$$

$$\mathbf{R}_{a,b} = \mathbf{R}_a + \exp[-\tau_a/\mu]\mathbf{U} + \mathbf{T}_a^*\mathbf{U}$$
(40)

$$T_{ab} = \exp[-\tau_b / \mu] \mathbf{D} + T_b \exp[-\tau_a / \mu_0] + T_b \mathbf{D}$$
(41)

In these equations, the product of two functions implies an integration over the appropriate solid angle so that all possible multiple scattering contributions are accounted for. For example:

$$\mathbf{R}_{a}^{*}\mathbf{R}_{b}(\mu,\phi;\mu',\phi') = \frac{1}{\pi} \int_{0}^{2\pi} \int_{0}^{1} R_{a}^{*}(\mu,\phi;\mu'',\phi'') \mathbf{R}_{b}(\mu'',\phi'';\mu',\phi') \mu'' d\mu'' d\phi''$$
(42)

The number M in Equation 37 is selected according to the convergence of the series, and varies from 5 to 12 in present calculations. The exponential terms in the adding equations are the direct transmission through layer a or b without scattering. The total transmission for the combined layer is the sum of the diffuse transmission  $T_{a,b}$  and the direct transmission  $T_{a,b}$  in the direction of the solar zenith angle  $\theta_0$ .

In the numerical computations, it is economical to set  $\tau_a = \tau_b$ . This is referred to as the doubling method. We start with an optical depth  $\tau \approx 10^{-8}$  and use Equations 31 through 34 to compute the reflection and transmission matrices. Equations 35 through 41 are subsequently employed to compute the reflection and transmission matrices for an optical depth of  $2\tau$ . The computations using these equations are repeated until the desired optical depth is obtained.

We divide the model vertical domain into 51 layers ( $\Delta p = 20$  mb for each layer except for the bottom layer, where  $\Delta p = 13$  mb). The doubling procedures are applied to each layer to obtain the layer reflection and transmission functions. Subsequently, the adding procedures are applied to the 51 layers to obtain the radiance at the top of atmosphere. For wavelengths between 3.5 and 5 µm, we take into account the thermal emission contributions in the solar flux transfer by adding the emission part  $(1 - \varpi)\pi B_V(T)$ , to the adding/doubling method in a manner described in Takano and Liou (1993), where  $\varpi$  is the single-scattering albedo and  $B_V(T)$  is the Planck function for a given layer temperature T. The thermal emission part, although small, has not been accounted for previously in broadband solar flux calculations and could be a significant energy source in the upper part of the atmosphere.

The input parameters required to drive the LBLE solar radiation model include the solar insolation; spectral band wave numbers of interest; solar and viewing zenith angles; the relative azimuthal angle; spectral surface albedos and emissivities; atmospheric temperature, humidity; and aerosol profiles; cloud phase; and cloud top and base altitudes. Figure 7 shows the flow diagram of the radiative transfer model developed for application to cloud remote sensing using VIIRS channel radiances. We first compile the spectral solar constant for 0.645, 1.6, 2.1, and 3.75µm VIIRS channels. The LBLE includes the options of using the detailed solar irradiance data given by Thekaekara (1976) (solar constant 1353 W/m²), Kneizys *et al.*, (LOWTRAN 7, 1988) (solar constant 1380 W/m²), and Anderson *et al.* (MODTRAN, 1995) averaged over appropriate spectral resolutions. We then generate the gaseous absorption line parameters from the updated HITRAN-1996 database for 0.645, 1.6, 2.25, 3.75 and 10.8 µm VIIRS channels.

### Radiative Transfer Flow Diagram For Remote Sensing Application

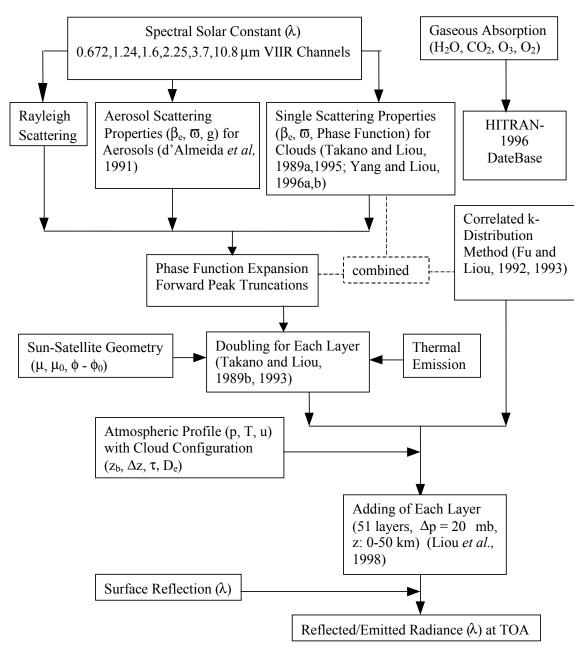


Figure 7. Flow diagram of the computation of TOA radiance for the application to cloud remote sensing using VIIRS channels.

The correlated k-coefficients for H<sub>2</sub>O covering the spectral region from 2,000 to 21,000 cm<sup>-1</sup> (0.5-5 µm) are determined based on the methodology developed by Fu and Liou (1992) in which a number of temperatures and pressures are used. The correlated k-coefficients for the 2.0 and 2.7 µm CO<sub>2</sub> bands are also derived, in which overlaps between H<sub>2</sub>O and CO<sub>2</sub> lines are accounted for by using the multiplication rule. Absorption due to O<sub>3</sub> and O<sub>2</sub> bands is also included in the model based on the Beer's Law (Liou, 1992). In addition, we have compiled the single-scattering properties of six typical aerosol types provided in d'Almeida *et al.* (1991) in connection with the LBLE model. In the present work, we use the rural aerosol model with 23 km visibility at the surface, which is the same as the default aerosol model in MODTRAN. At this point, the entire solar spectrum is divided into a total of 380 intervals, each of which is 50 cm<sup>-1</sup> wide. For each spectral interval, the inverse of the cumulative probability function k(g) is evaluated at 30 g values, where 0 < g < 1.

We then compute single-scattering parameters of air molecules, aerosol and cloud particles. We also construct the table of correlated-*k* coefficients. The resulting single-scattering parameters, cumulative *k*-distribution functions, phase functions, and auxiliary data were combined as described in section 3.3.1.6, and input into the radiative transfer program.

### Intercomparisons with results from MODTRAN

Over the years, a hierachy of fast radiative transfer models have been developed by the Air Force Research Laboratory (previously the Air Force Geophysics Laboratory), including LOWTRAN 5-7 (Kneizys *et al.*, 1988; Isaacs *et al.*, 1986, 1987), FASCODE and MODTRAN. At present, MODTRAN is the most advanced narrow-band radiative transfer program for computing transmittances in cloudless atmospheres. MODTRAN covers the spectral range from 0 to 50,000 cm<sup>-1</sup> (~0.2 - ∞ μm), with flexible spectral resolution between 50 cm<sup>-1</sup> and 2 cm<sup>-1</sup> (Berk *et al.*, 1989). The latest version of MODTRAN (version 3.7) maintains the capability of the older versions. Several atmospheric temperature, humidity, ozone and aerosol (haze) profiles are built into the radiative transfer model. The default atmospherice profiles include the tropical, mid-latitude summer, mid-latitude winter, sub-arctic summer, sub-arctic winter, and the 1976 US Standard atmosphere. The default aerosol models include rural, maritime, urban , tropospheric (humidity dependent), and desert (wind dependent) models. Band model parameters based on HITRAN line data for twelve atmospheric molecular species (H<sub>2</sub>O, CO<sub>2</sub>, O<sub>3</sub>, N<sub>2</sub>O, CO, CH<sub>4</sub>, O<sub>2</sub>, NO, SO<sub>2</sub>, NO<sub>2</sub>, NH<sub>3</sub>, and HNO<sub>3</sub>) are included in the model (Kneizys *et al.*, 1988).

Larsen (1994) replaced the two-stream multiple scattering computations in the MODTRAN with a modified version of the Discrete Ordinate Radiative Transfer (DISORT) Model. DISORT is a state-of-the-art, user-friendly discrete ordinate algorithm for radiative transfer in vertically inhomogeneous, non-isothermal, plane-parallel media (Stamnes *et al.*, 1988). It considers scattering, absorption, emission by particles as well as incident radiative sources at boundaries. Given cloud optical properties for each layer, DISORT can generate angular dependent radiance and radiative flux profiles within the cloud.

Although MODTRAN can produce spectrally high-resolution clear radiance with excellent accuracy, and DISORT can generate reliable cloud radiance at any angle, the present combination of MODTRAN with DISORT (MD) is less than ideal. In the original MODTRAN multiple scattering program, *azimuthally independent* source functions are computed by the two-stream method. These source functions are then substituted into the integrated form of the radiative transfer equation to obtain the radiance distribution. Thus the resulting radiance distribution is also *azimuthally* 

independent. The MD program still follows this approach. That is, instead of using source functions from the two-stream radiative transfer scheme built into the MODTRAN, MD uses DISORT-generated azimuthally independent source functions to compute azimuthally independent radiance distribution. Effects of this model artifact can be seen from the following comparison of clear and cloudy radiances from the LBLE and the MODTRAN/MD.

To compare the performance of the LBLE with the MD, we first compare the clear radiance generated by the two models. Figure 8 shows the geometry of the sun and the sensor with respect to the target along with the specification of angular parameters. In the planned NPOESS afternoon orbit, the sun is always to the west of the target, while the VIIRS sensor can be either to the east or to the west of the target. The angular parameters can be derived from simulated orbital information.

### Sun-Satellite Geometry

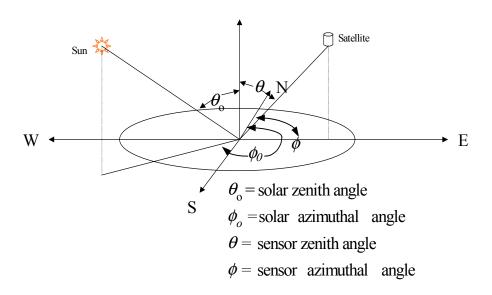


Figure 8. Sun satellite geometry.

Table 3 shows the comparison of the 0.645  $\mu m$  clear and haze radiance from the LBLE model (denoted as UCLA) and from the MODTRAN/MD model for various sun-sensor geometries, including high-sun, low-sun, nadir-sensor, and off-nadir-sensor configurations. The terms "clear" and "haze" mean without and with effects of background aerosols, respectively. In general, radiance from the LBLE model agrees with the MD radiance. For the high-sun cases (1, 3, 5), the differences are less than 10 percent. For the low-sun cases (2,4,6), the differences are larger but are still less than 15 percent. When the sun is low, reflection due to aerosol scattering is stronger, so that the error in the azimuthally independent MD radiance increases. Similarly, for the same solar zenith angle, the differences for the off-nadir cases (3-6) is generally larger than those for the nadir case (1,2) due to errors caused by the assumption of azimuthally independent scattering in the MD calculations. For the off-nadir cases, the differences are smaller for the sensor at forward-scattering angle ( $\Delta \phi > 90^{\circ}$ ) than at back-scattering angle ( $\Delta \phi < 90^{\circ}$ ).

Table 3. Comparison of the 0.645 µmclear and haze radiance from the LBLE model (UCLA) and the MODTRAN/MODTRAN-DISORT model for various sun-sensor geometries, including high-sun, low-sun, nadir-sensor, and off-nadir-sensor configurations.

Case No.	Sky		Clear		Haze		
1.0.		UCLA	MODTRAN	% Diff.	UCLA	MODTRAN	% Diff.
1	$\theta_{o} = 32^{o}$ $\theta = 0^{o}$	1.44200	1.39506 (1.40846) <sup>†</sup> [1.34247]@	-3.25 (-2.33) [-6.90]	1.67600	1.67338 (1.72751) [1.63477]	-0.16 (3.07) [-2.46]
2	$\theta_{o} = 75^{o}$ $\theta = 0^{o}$	0.51353	0.49331 (0.50555)	-3.94 (-1.55)	0.80821	0.72738 (0.79451)	-10.00 (-1.70)
3	$\theta_{o} = 32^{o}$ $\theta = 40^{o}$ $\approx \phi = 136^{o}$	1.57441*	1.481337	-5.91	1.88152	1.92653	2.39
4	$\theta_o = 75^{\circ}$ $\theta = 40^{\circ}$ $\cong \phi = 168^{\circ}$	0.735682	0.677275	-7.93	0.976508	1.059228	8.47
5	$\theta_{o} = 32^{o}$ $\theta = 40^{o}$ $\approx \phi = 0^{o}$	1.36383	1.302619	-4.49	1.92005	1.756079	-8.54
6	$\theta_{o} = 75^{\circ}$ $\theta = 40^{\circ}$ $\cong \phi = 0^{\circ}$	0.612285	0.596909	-2.51	1.67525	1.4494	-13.48

<sup>\*:</sup> in W/m²/sr; @: Numbers in brackets are based on MODTRAN results from TASC

Table 4 shows the comparison of the 3.75 µm clear and haze radiance from the LBLE and the MODTRAN/MD model for various sun-sensor geometries, including high-sun, low-sun, and nadir-sensor configurations. The differences in the solar component are less than 15 percent. The differences in the thermal IR component is less than 0.4 percent. For the total radiance, the differences are less than 4 percent, because the IR component dominates the total radiance. In addition, since the fraction of the solar component in the total radiance decreases as the solar zenith angle increases, the total radiance differences are less for the low-sun case than for the high-sun case. Finally, Table 5 shows the comparison of 10.8 µm clear and haze radiances from the LBLE model and the MODTRAN/MD model for the nadir-sensor case. The differences are less than 5 percent for both clear and haze cases. Overall, it is demonstrated that the LBLE clear radiance compare reasonably well with the MD radiance, though small errors due to neglecting azimuthal dependence in the computation of aerosol scattering component do exist.

<sup>†:</sup> Numbers in parentheses are based on DISORT 8-stream source functions

Table 4. Comparison of the 3.75 µm clear and haze radiance from the LBLE model (UCLA) and the MODTRAN/MODTRAN-DISORT model for various sun-sensor geometries, including high-sun, low-sun, and nadir-sensor configurations.

Co mp one nt	Case No.	Sky	Clear			Haze			
			UCLA	MODTRAN	% Diff.	UCLA	MODTRAN	% Diff.	
Sol	1	$\theta_{o} = 32^{o}$ $\theta = 0^{o}$	0.02327*	0.02140 ( 0.02140)†	-8.04 (-8.04)	0.02331	0.02170 (0.02610)	-6.91 11.97	
ar	2	$\theta_{o} = 75^{o}$ $\theta = 0^{o}$	0.006321	0.005400 (0.00540)	-14.57 (-14.57)	0.006548	0.005600 (0.005600)	-14.48 -14.48	
Th er	3	$\theta_{o} = 32^{o}$ $\theta = 0^{o}$	0.04354	0.04340 ( 0.04350)	-0.32 ( -0.09 )	0.04344	0.04330 (0.04330) [ 0.04321]	-0.32 (-0.32) -0.53	
mal IR	4	$\theta_{o} = 75^{o}$ $\theta = 0^{o}$	0.04354	0.04340 (0.04350)	-0.32 (-0.09)	0.04344	0.04330 (0.04330)	-0.32 (-0.32)	
Tot al	5	$\theta_{o} = 32^{o}$ $\theta = 0^{o}$	0.06681	0.06480 ( 0.06490) [0.064469]	-3.01 -2.86 [-3.50]	0.06675	0.06500 (0.06940) [0.064429]	-2.62 3.97 [-3.48]	
	6	$\theta_{o} = 75^{o}$ $\theta = 0^{o}$	0.04986	0.04940 (0.04890)	-0.92 ( -1.93)	0.04999	0.04890 (0.04890)	-2.18 (-2.18)	

<sup>\*:</sup> in W/m 2/sr

Table 5. Comparison of the 10.8 µm clear and haze radiance from the LBLE model (UCLA) and the MODTRAN/MODTRAN-DISORT model for various sensor geometries, including nadir-sensor and off-nadir-sensor configurations.

Case		Clear		Haze			
No.	UCLA	MODTRAN	% Diff.	UCLA	MODTRAN	% Diff.	
1 (θ = 0°)	7.90516	7.56000	-4.37	7.88450	7.5489	-4.26	
2 (θ = 40 °)	7.86735			7.83776			

<sup>†:</sup> Numbers in parentheses are based on DISORT 8-Stream source functions

We then compare the cirrus cloudy radiance generated by the two models. The 1976 US Standard Atmosphere is assumed. A cirrus cloud is prescribed to be at 9-10km. Figure 9 shows the contour plot of the 0.645 µm reflectance and radiance for various combinations of cirrus cloud optical depths and mean effective sizes and for a high-sun angle ( $\theta_0 = 32^{\circ}$ ) and nadir-sensor. It is obvious that the radiance/reflectance depends strongly on optical depth but weakly on the mean effective size. For optical depths between 0 and 8, the reflectance varies between 0 and 0.5, and the radiance varies between 0 and 12 W m<sup>-2</sup> sr<sup>-1</sup>. The computed range of radiance can help the determination of maximum and minimum values in the design of the VIIRS sensor measurement range. Figure 10 shows the contour plot of the 3.75 µm total (solar + thermal) reflectance and radiance for the same combination of cirrus cloud optical depths and mean effective sizes and for a high-sun angle ( $\theta_0$  = 32°) and nadir-sensor. For small mean effective sizes, the radiance/reflectance is virtually independent of optical depth, but for larger mean effective sizes, the radiance/reflectance is nearly independent of mean effective size. For mean effective size less than 40 µm, the total radiance varies between 0.04 and 0.08 W m<sup>-2</sup> sr<sup>-1</sup>. For mean effective size larger than 40 µm, the total radiance varies between 0.01 and 0.06 W m<sup>-2</sup> sr<sup>-1</sup>. Figure 11 shows the contour plot of the 0.645 and 3.75 µm radiance percentage differences between LBLE and MD results.

For the 0.645  $\mu$ m radiance, differences are generally independent of mean effective sizes, but increase with increasing optical depth. For optical depths greater than 8, the differences exceed 40 percent. For the 3.75  $\mu$ m radiance, differences are less than 10 percent for optical depths less than 2. However, for mean effective size larger than 50  $\mu$ m, the differences increase with the optical depth. Figure 12 shows the contour plots of the 10.8  $\mu$ m MD radiance and the differences between LBLE and MD radiance. The radiance decreases with increasing optical depth. There is a weak dependence on the mean effective size. As in the cases of 0.645  $\mu$ m and 3.75  $\mu$ m, for optical depths greater than 2, the differences increase with optical depth up to about 30 percent. In summary, the above results of intercomparison of cirrus cloudy radiance shows that MD radiances for optically thick clouds differ from the LBLE radiances by more than 10 percent in general. Although such inter-comparisons have been done only for one cirrus cloud scenario, we expect that similar patterns of differences between LBLE and MD radiance will show up for other cirrus and water cloud scenarios. The inherent problem of neglecting the azimuthal dependence of radiance will cause the resulting radiance for optically thick clouds to be in large error in every case.

The above intercomparison of LBLE and MD results leads to the following conclusions. The MD, or a program based on a similar philosophy of assuming azimuthally independent scattering, is not feasible to be used as the radiative transfer model for the development of retrieval algorithms for inferring cloud optical depth and effective particle size. This is simply because the MD is defective in the treatment of multiple scattering effect and has not been used in any previous algorithm development of retrieval of cloud optical and microphysical parameters. In addition, because the development of the present retrieval programs is closely coupled with the physically based LBLE, retrievals of cloud optical depth and effective particle size are expected to meet the EDR threshold requirement only if radiances generated by LBLE are used as input. In fact, having recognized that retrievals of the cloud parameters must be properly coupled with a reliable forward radiative transfer program, established cloud retrieval algorithm development teams (e.g., King *et al.* at NASA Goddard Space Flight Center, Minnis *et al.*, at NASA Langley Research Center, and Rossow *et al.* at NASA Goddard Institute of Space Science) all have developed their own radiative transfer programs based on sound physical principles for the purpose of algorithm development and sensitivity studies. Therefore, it appears both scientifically justified and commercially advantageous that the LBLE

should be used as the only radiative transfer model for the development of retrieval algorithms for inferring cloud effective particle sizes and optical depths.

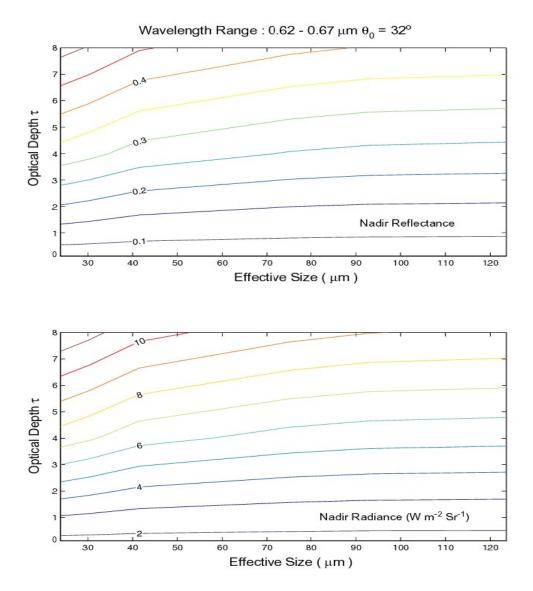


Figure 9. The contour plot of the 0.645  $\mu m$  reflectance and radiance for various combinations of cirrus cloud optical depths and mean effective sizes and for high-sun angle ( $\theta_0 = 32^\circ$ ) sensor at nadir angle.

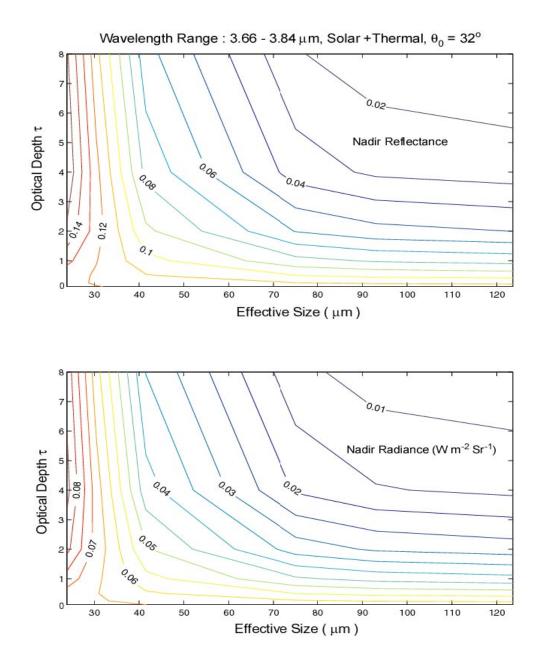


Figure 10. The contour plot of the 3.75  $\mu$ m total (solar + thermal) reflectance and radiance for the same combination of cirrus cloud optical depths and mean effective sizes and for high-sun angle ( $\theta_0 = 32^\circ$ ) sensor at nadir angle.

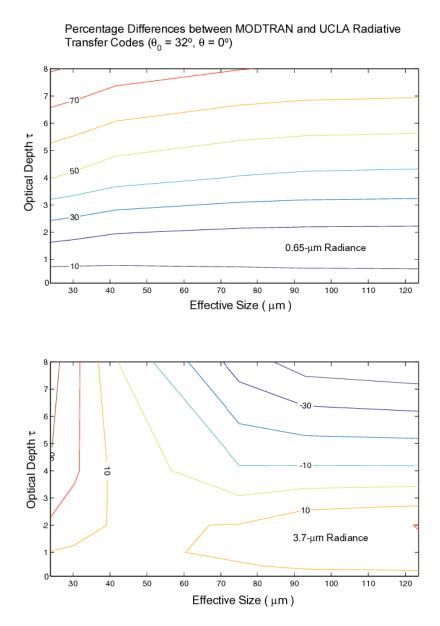


Figure 11. The contour plot of the 0.645 and 3.75  $\mu m$  radiance percentage differences between LBLE and MODTRAN-DISORT.

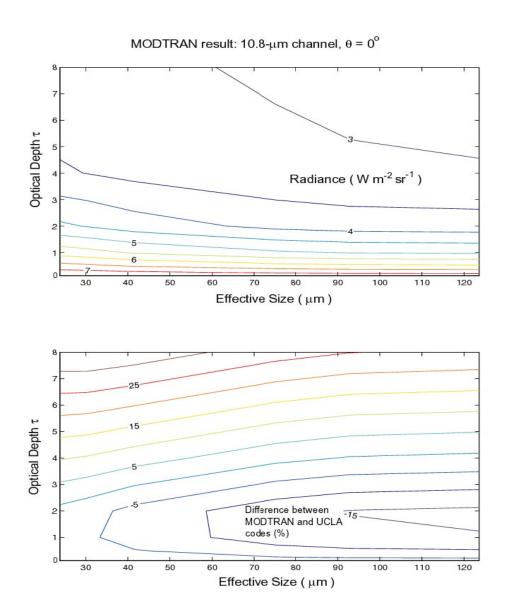


Figure 12. The contour plots of the 10.8 µm MODTRAN-DISORT radiance and the differences between LBLE and MODTRAN-DISORT radiance.

### 3.3.2 Mathematical Description of the Algorithms

### 3.3.2.1 Retrieval of Cirrus Cloud Parameters

Figure 13 shows the flow chart of cirrus cloud retrieval program. Once the cirrus cloud pixels are identified, three retrieval programs are applied to nighttime and daytime data. For the nighttime retrieval, the IR cloud retrieval program (C1) using the VIIRS 3.75 and 10.8 μm radiance is employed to infer the cirrus cloud temperature, mean effective size, and visible optical depth. For the daytime retrieval, two complementary programs are used. The solar cloud retrieval algorithm (C2) using the VIIRS 0.645, 1.6, and 2.13 μm radiance is employed to infer cirrus cloud optical depth and mean effective size. The daytime IR cloud retrieval algorithm (C3) based on the same retrieval principle as algorithm C1 is employed to infer daytime cirrus cloud temperature, mean effective size, and visible optical depth. However, the 3.75 μm radiance for local daytime contain solar radiation that is reflected by the earth-atmospheric system. Thus it is necessary to remove the 3.75 μm solar component for the daytime IR algorithm to be applicable.

## Cirrus Cloud Retrieval Program

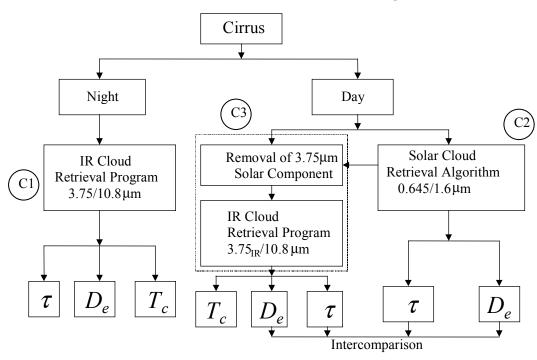


Figure 13. Flow chart for the cirrus cloud retrieval program

The retrieval program for deriving cirrus temperature, mean effective crystal size, and optical depth from the upwelling radiance of VIIRS cloud retrieval channels follows the principles of the dual-IR-channel technique presented in Liou *et al.* (1990). The VIIRS 3.75 µm and 10.8 µm radiances have been selected for the development. A major advantage of using these two channels for cirrus retrievals is that the radiance of these window bands is affected little by the presence of water vapor.

From the theory of radiative transfer, we may express the upwelling radiance at TOA for the 3.75 and 10.8  $\mu$ m channels over a cirrus cloudy atmosphere in terms of the cirrus cloud-top temperature  $T_c$  and emissivities  $\varepsilon_{1,2}$  as follows:

$$R_i = (1 - \varepsilon_i) R_{ai} + \varepsilon_i B_i(T_c) \quad i = 1, 2 \tag{43}$$

where  $R_{ai}$  denote the upwelling radiance reaching the cloud base for the two spectral bands and  $B_i(T_c)$  are the respective Planck functions at  $T_c$ . The first term on the right-hand side of Equation 43 represents the contribution of the transmitted radiance from below the cloud. The second term denotes the emission contribution from the cloud itself. The emission by water vapor above the cirrus cloud has been neglected. The effects of cloud reflectivity, which are generally less than 3 percent of the incident radiance based on exact radiative transfer calculations, have also been neglected.

In order to solve Equation 43 for  $T_c$  and  $\varepsilon_i$  numerically, we need to correlate  $\varepsilon_1$  and  $\varepsilon_2$ , and express  $B_1(T_c)$  in terms of  $B_2(T_c)$ . The clear radiance  $R_{a1, 2}$  must also be known. First we compute the Planck functions  $B_1(T_c)$  and  $B_2(T_c)$ , taking into account the filter functions of both channels. A look-up table for both  $B_1(T_c)$  and  $B_2(T_c)$  in 0.1 K intervals is constructed by using a range of  $T_c$  from 150 to 300 K. Values in the look-up table are then fitted to a third-degree polynomial based on a least-square regression technique:

$$B_1(T_c) = \sum_{n=0}^{3} a_n [B_2(T_c)]^n = f[B_2(T_c)]$$
(44)

where the coefficients are  $a_0 = 2.6327 \times 10^{-4}$ ,  $a_1 = -1.063 \times 10^{-4}$ ,  $a_2 = 8.2976 \times 10^{-6}$ , and  $a_3 = 3.7311 \times 10^{-7}$ . Errors in the fitted polynomial are less than 1 percent.

Second, we investigate the relationship between the emissivities for the two channels. From radiative transfer calculations, and following the approach proposed by Liou *et al.* (1990), we may parameterize cirrus emissivities at the 3.75 and 10.8  $\mu$ m wavelengths in terms of visible optical depths  $\tau$  in the form:

$$\varepsilon_i = 1 - \exp(-k_i \tau) \qquad i = 1, 2 \tag{45}$$

The exponential term represents the effective transmissivity. The parameters  $k_1$  and  $k_2$  represent the effective extinction coefficients for the two channels, accounting for multiple scattering within cirrus clouds and for the difference between visible and IR extinction coefficients. Both  $k_1$  and  $k_2$  are smaller than 1 because the effect of multiple scattering increases transmission. Thus the products  $k_i\tau$  may be considered as effective optical depth that would yield the same emissivity values for the non-scattering conditions at the 3.75 and 10.8  $\mu$ m wavelengths. By eliminating  $\tau$  from Equation 44, we obtain

$$(1 - \varepsilon_1)^{\frac{1}{k_1}} = (1 - \varepsilon_2)^{\frac{1}{k_2}} \tag{46}$$

Equation 46 correlates  $\varepsilon_1$  and  $\varepsilon_2$  directly. A further combination of Equations 42 and 45 leads to:

$$\left[ (R_1 - B_1(T_c)) / (R_{a1} - B_1(T_c)) \right]^{\frac{1}{k_1}} = \left[ (R_2 - B_2(T_c)) / R_{a2} - B_2(T_c) \right]^{\frac{1}{k_2}}$$
(47)

Substitution of Equation 43 into Equation 46 results in a nonlinear algebraic equaiton, with  $B_1(T_c)$  as the only unknown:

$$[(R_2 - B_2(T_c))/(R_{a2} - B_2(T_c))] - [(R_1 - f[B_2(T_c)])/(R_{a1} - f[B_2(T_c)])]^{k_2/k_1} = 0$$
(48)

We investigate the dependence of  $k_2/k_1$  on the particle size distribution and optical depth based on radiative transfer calculations by using the six ice crystal size distributions given in Section 3.3.1.1.1. A reasonable range of optical depth for these size distributions has been used. As an initial approximation, we may take  $k_1$  and  $k_2$  as independent of the optical depth. However, for more accurate retrieval of cirrus cloud parameters, we need to consider the dependence of  $k_2/k_1$  on the optical depth. Generally,  $k_2/k_1$  decreases as  $D_e$  increases. For a small  $D_e$  (~20 µm),  $k_2/k_1$  is close to 2. This is primarily because the single-scattering albedo is larger for the 3.75 µm wavelength (~0.8) than that for the 10.8 µm wavelength (~0.4), which implies that more scattering is associated with the former wavelength. For  $D_e > 100$  µm,  $k_2/k_1$  approaches 1 for the following reasons: First, the extinction coefficients are approximately the same for the two wavelengths because of large-size parameters in which the geometric optics limit is valid. Second, the single-scattering albedos are also approximately the same for these wavelengths because substantial absorption occurs within large ice crystals. This implies that only the diffracted and externally reflected light contributes to the scattering processes. For obtaining the initial guesses,  $k_2/k_1$  can be expressed in terms of  $1/D_e$  in the form:

$$k_2 / k_1 = \sum_{n=0}^{2} b_n D_e^{-n} \tag{49}$$

where the coefficients  $b_n$  are determined from a second-order least square method using the radiance tables. In the retrieval scheme, however,  $b_n$  are further expressed as third-order polynomial functions of  $\tau$ :

$$b_n(\tau) = \sum_{n=0}^{3} d_n \tau^n \tag{50}$$

where the coefficients  $d_n$  are obtained from a third-order least square method.

A direct determination of  $D_e$  from the data that are available from the present satellite thermal infrared radiometers appears to be very difficult, if not impossible. However, we may relate  $D_e$  to the cloud temperature through appropriate observations. Based on a large number of cirrus microphysical data collected by optical probes during flights over midlatitudes. Heymsfield and Platt (1984) have suggested that ice crystal size distribution can be represented by a general power form as follows:

$$n(L) = \begin{cases} A_1 L^{b_1}(IWC) & L \le L_0 \\ A_2 L^{b_2}(IWC) & L > L_0 \end{cases}$$

$$(51)$$

where  $L_0 = (A_2/A_1)^{\frac{1}{b_1-b_2}}$ , *IWC* is the ice water content, and  $A_{1,2}$  and  $b_{1,2}$  are temperature dependent empirical coefficients determined from the measured data. Heymsfield and Platt (1984) have shown that for a given temperature, values of  $A_{1,2}$ ,  $b_{1,2}$ , and *IWC* may be parameterized in terms of

temperature in the range of  $-20^{\circ}$  to  $-60^{\circ}$ C (Liou, 1992). Based on this parameterization, the mean n(L) is also a function of temperature. Moreover, aircraft and laboratory measurements indicate that the width D and the length L of a hexagonal ice crystal are related (e.g. see Auer and Veal, 1970) in which a parameterization relation can be developed for the two. Thus, with the functional form of n(L) determined, the temperature-dependent mean effective size  $< D_e >$  can be obtained, where the bracket denotes the mean value for a given temperature. Subsequently, we perform a least squares polynomial fitting to relate  $< D_e >$  to  $T_c$  in the form:

$$\langle D_e \rangle = \sum_{n=0}^{3} c_n (T_c - 273)^n$$
 (52)

where  $c_0 = 326.3$ ,  $c_1 = 12.42$ ,  $c_2 = 0.197$ , and  $c_3 = 0.0012$ . Equation 52 was used in our previous retrievals using the AVHRR channel data (Ou *et al.*, 1993; Rao *et al.*, 1995).

The retrieved  $D_e$  based on Equation 52 may be in error due to uncertainties in  $A_{1,2}$ ,  $b_{1,2}$ , and IWC. To improve the accuracy of the retrieved mean effective size, we modify the value of  $D_e$  according to the following procedures. Let  $D_e$  be the modified value of  $D_e$ , we assume that  $D_e$  and  $D_e$  have the following relationship based on dimensional analysis:

$$D_{a} \propto (IWC)^{\frac{1}{3}} \tag{53}$$

Observational evidence (Heymsfield and Platt, 1984) shows that:

$$\langle D_e \rangle \propto \langle (IWC) \rangle^{1/3} \tag{54}$$

where  $\langle IWC \rangle$  is the temperature-dependent mean value of IWC, derived from observations. According to Liou (1992),  $\langle IWC \rangle$  can be parameterized in terms of  $T_c$  as follows:

$$<$$
IWC>=exp{-7.6+4 exp[-0.2443x10<sup>-3</sup>(253-T<sub>c</sub>)]<sup>-2.445</sup>}  
For  $T_c < 253$  K (55)

The quantity *IWC* may be expressed by (Fu and Liou, 1993):

$$IWC = \tau [\Delta z(\alpha + \beta/D_e)]$$
 (56)

where  $\tau$  is the visible optical depth,  $\Delta z$  is the cloud thickness, and  $\alpha$  and  $\beta$  are empirical constants. Combining Equations 52 through 56, we obtain:

$$D_e = c \left\{ \pi \left[ \Delta z(\alpha + \beta/D_e) < IWC > \right] \right\}^{1/3} < D_e >$$
(57)

where c is a proportional factor. Equation 57 is an implicit algebraic equation for  $D_e$ , and it replaces Equation 52 as the closure equation for the retrieval problem. It has been verified using the balloon-borne replicator data collected on November 26 and December 5, 1991 during the FIRE-II IFO (Ou *et al.*, 1995).

Finally, in order to solve Equation 48 for  $B_2(T_c)$ , the upwelling radiance reaching the cloud base,  $R_{a1,2}$ , must be given. We may approximate these radiance values by the radiance measured from satellites in

clear conditions because little water vapor is present above cirrus clouds. In the case of VIIRS data, each scan line spans ~3000 km. Because of the high resolution of the VIIRS pixels, it is most likely that, over a large area scanned by the radiometer, some of the data points correspond to clear conditions. It follows that if we can identify the clear pixels,  $R_{a1,2}$  may be determined by a statistical method. We may select a scene and use the data within this scene to construct a two-dimensional histogram in the domain of  $R_1$  and  $R_2$ . The area of the scene should be large enough to contain a statistically significant amount of pixels, but, at the same time, this area should also be sufficiently small to guarantee the homogeneity of the surface temperature and water vapor distributions within the scene. Usually, a 1° by 1° scene is adequate for the analysis. The radiance corresponding to the peak of the frequency distribution is assigned as the mean clear radiance. In the algorithm sensitivity studies, the clear radiance is pre-determined from the LBLE program.

Thus Equations 43 (2 equations), 44, 45 (2 equations), 49, and 57 form a complete set of seven governing equations for the solution of seven unknowns:  $\varepsilon_{1,2}$ ,  $B_{1,2}(T_c)$ ,  $T_c$ ,  $\tau$ , and  $D_e$ . Figure 14 shows a schematic description of the solution procedures. Using the VIIRS 3.7 and 10.8 µm channel data, we first obtain an estimate of mean cirrus cloud temperature from an optimization scheme (Appendix B). Subsequently, an initial guess of  $\varepsilon_i$ ,  $\tau$ ,  $k_2/k_1$ ,  $D_e$  and the proportional factor c (Equation 56) can be computed from the following equations:

$$\hat{\varepsilon}_2 = \frac{R_{ai} - R_2}{R_{ai} - B_2(\hat{T}_c)} \tag{58}$$

$$\hat{\tau} = -\ln(1 - \hat{\varepsilon}_2)/\hat{k}_2 \tag{59}$$

$$\hat{k}_2 / \hat{k}_1 = \frac{\ln[R_2 - B_2(\hat{T}_c)] / \ln[R_{a2} - B_2(\hat{T}_c)]}{\ln[R_1 - B_1(\hat{T}_c)] / \ln[R_{a1} - B_1(\hat{T}_c)]}$$
(60)

$$\hat{D}_e = \frac{2b_2}{-b_1 + \sqrt{b_1^2 - 4b_1(b_0 - \hat{k}_2/\hat{k}_1)}}$$
(61)

$$c = \frac{\hat{D}_e}{\left\{\hat{\tau}/\left[\Delta z(\alpha + \beta/\hat{D}_e) < IWC >\right]^{1/3} < D_e > \right\}}$$
(62)

where Equations 58 through 62 are the inverse form of Equations 43, 45, 48, 49, and 57. For the above initial guesses, we prescribe  $k_2$  as constant, although it is a function of mean effective size. Utilizing the estimated value of  $k_2/k_1$ , we proceed to solve Equation 57 for  $T_c$  by using a modified Newton's iteration scheme, which is efficient for the solution of nonlinear algebraic equations. Subsequently, we use Equations 43, 45, 49, 50, and 57 to obtain new estimates of  $\varepsilon_i$ ,  $\tau$ ,  $k_2/k_1$ , and  $D_e$ . To solve Equation 57, we first determine <IWC> and  $<D_e>$  from Equations 55 and 52, respectively, using the retrieved cloud temperatures. Subsequently, we solve for  $D_e$  numerically based on the retrieved value of  $\tau$  and the prescribed value of  $\Delta z$ . Then, we put the new estimate of  $T_c$ ,  $\tau$ , and  $D_e$  to convergence tests. If the differences between the new estimate of these parameters and that for the previous iteration are less than prescribed convergence thresholds, then the new estimate of  $T_c$ ,  $\tau$ , and

 $D_e$  is the solution. Otherwise, we go back to solve Equation 48 using the new estimate of  $k_2/k_1$ , and repeat the subsequent processes.

## Flow Chart for Cirrus Nighttime IR Retrieval Algorithm

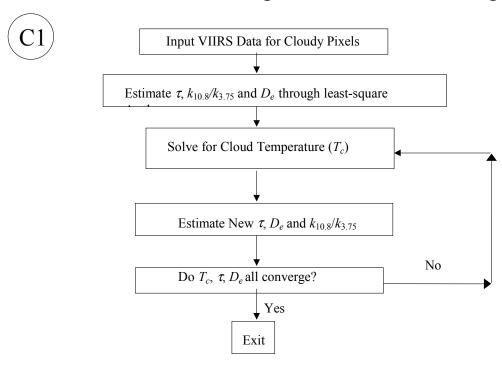


Figure 14. Flow chart for nighttime IR cirrus retrieval algorithm.

### Solar cirrus cloud retrieval program

A generalized schematic description of the solar cirrus cloud retrieval algorithm is given as follows. For the daytime retrieval, the solar cloud retrieval program (C3) using the VIIRS 0.645, 1.24, 1.6 and 2.13 µm channel radiance is utilized to infer the cirrus cloud optical depth and mean effective size. The retrieval program follows the principles of the dual-solar-channel technique presented in Nakajima and King (1990). Figure 15 illustrates the underlying principles behind the simultaneous determination of cirrus cloud optical depth and mean effective size from reflected solar radiation measurements. Preliminary calculations using the MAS 0.681, 1.617, and 2.139 µm channels have been carried out. Radiative transfer calculations were performed for the six ice crystal size distributions described in section 3.3.1.1 using the LBLE. The results are displayed in twodimensional reflectance diagrams (0.681 - 1.617 µm and 0.681 - 2.139 µm) for optical depths ranging between 0.5 and 64. Also shown are the results for water clouds having mean effective radii of 4 and 8 um, along with the MAS data obtained from FIRE-II-IFO on December 5, 1991. The top and bottom diagrams correspond to the cases for water and land surfaces, respectively. In the calculations, the effective surface albedos used were determined from the MAS reflectances over clear pixels that were identified from the scheme similar to the one developed by Ou et al. (1996). See section 3.3.2.1. for further discussions.

The preliminary calculations illustrate the information content of optical depth and mean effective size in the MAS 0.681, 1.617, and 2.139  $\mu$ m reflectances. The nearly horizontal or diagonal curves are constant- $D_e$  contours for the six ice crystal size distributions, while the nearly vertical dotted lines are the constant- $\tau$  contours for the selected optical depths. The underlying surface was assumed to be Lambertian. The minimum value of the reflection function at each wavelength corresponds to the reflection function of the underlying surface at that wavelength in the absence of an atmosphere. The 0.681 and 1.617  $\mu$ m reflectance correlation shows that if the mean effective radius of a water cloud (stratus and cumulus) is less than about 10  $\mu$ m and if the mean effective size of an ice cloud is larger than about 20  $\mu$ m, a clear distinction can be made between the two in the correlation domain. Also noted is that the 0.681  $\mu$ m reflectance depends more strongly on optical depth than on particles size. However, the 1.617 and 2.139  $\mu$ m reflectances are primarily functions of particle size rather than optical depth. From the diagrams, as the optical depth increases, the constant- $D_e$  contour lines gradually approach asymptotic values. Consequently, these three channels can be utilized to determine both optical depth and particle size.

The MAS data were obtained from the flight missions during FIRE-II IFO. The case over water was taken at 1636 UTC, 5 December 1991, when ER-2 was flying over the northern part of the Gulf of Mexico near the southern coastal region of Louisiana. The case over land was taken at 1923 UTC, 5 December 1991, when ER-2 was flying over eastern Oklahoma. The data points indicate that the detected cirrus clouds appear to contain small ice particles with optical depths less than about 6. Larger optical depths indicate the possibility of cirrus overlying low clouds.

Figure 16 shows an algorithm for retrieving the optical depth and mean effective size of cirrus clouds from comparisons of measured reflection functions with entries in the library. We first prescribe the sun-sensor geometric parameters, including the solar zenith angle, the sensor viewing zenith angle, and the relative azimuthal angle for each pixel. In the application of this scheme to real-time satellite data, the surface albedo can be obtained from the histogram of clear reflectances. In the algorithm sensitivity studies, the surface albedo is determined based on climatological data sets. We then select values for the mean effective ice crystal sizes and optical depths. Subsequently, we construct radiance tables for each combination of ice crystal mean effective size and optical depth and for each VIIRS channel used for cloud retrieval (0.645, 1.6, and 2.13 µm). The computation of the reflection function in the radiance table has been described in section 3.3.1.3. A series of numerical iteration procedures are set up to search for the simulated reflectances that best match the measured reflectances (Appendix C). In the algorithm sensitivity studies, the "measured" reflectances are precomputed using the LBLE. The numerical scheme is basically composed of minimizing the sum of the square of the "residual" for each channel. The "residual" is defined as the difference between the logarithm of simulated reflectance and the logarithm of the measured reflectance. This definition of the residual used for determining the best fit is typically defined as a least-squares fit (Twomey and Cocks, 1989). The retrieved cloud parameters consist of the combination of optical depth and effective particle size that is associated with the best-fit reflectances, yielding the minimal residual.

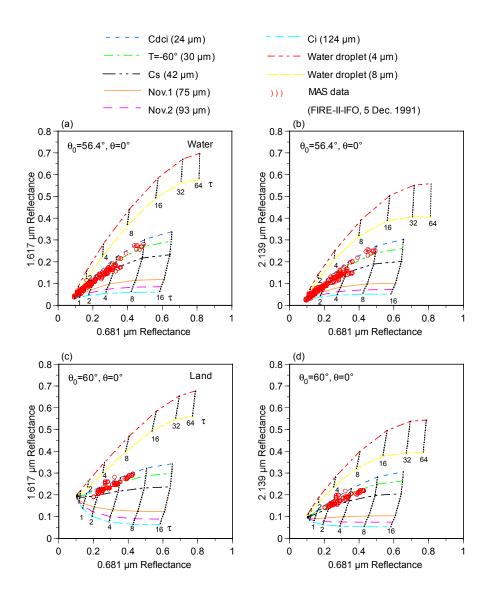


Figure 15. MAS data taken from FIRE-II IFO over both land and water surfaces. The plane-parallel radiative transfer calculations were performed for the six ice crystal size distributions presented in section 3.3.1.1 and for two water clouds with mean effective radii of 4 and 8  $\mu m$ . The optical depth ranges from 0.5 to 64.

### Flow Chart for Cirrus Solar Retrieval Algorithm Determine Geometry and Surface Albedo for each Pixel: VIIRS & Near-IR Radiance Data $\mu$ , $\mu_0$ , $\phi$ , $Ag(\lambda)$ Calculated libraries for $D_e = D_{e1}, D_{e2}, D_{e3}, ..., D_{en}$ $\tau = \tau_1, \tau_2, \tau_3, \ldots, \tau_m$ Measured Reflectances Compute Reflectance Functions $r_{meas}(\lambda=0.645 \,\mu\text{m}, \,\mu, \,\mu_0, \,\phi)$ $r_{comp}(\lambda=0.645 \,\mu\text{m}, \,\tau_m, D_{en}; \mu, \mu_0, \,\phi)$ $r_{meas}(\lambda=1.615\mu\text{m}, \mu, \mu_0, \phi)$ $r_{comp}(\lambda=1.615\mu\text{m}, \tau_m, D_{en}; \mu, \mu_0, \phi)$ $r_{meas}(\lambda=2.132\mu\text{m}, \mu, \mu_0, \phi)$ $r_{comp}(\lambda=2.132\mu\text{m}, \tau_m, D_{en}, \mu, \mu_0, \phi)$ Residual Definition: $\{ln[r_{comp}(\lambda=0.645\mu\text{m}, \tau_m, D_{en}; \mu, \mu_0, \phi)] - ln[r_{meas}(\lambda=0.645\mu\text{m}, \mu, \mu_0, \phi)]\}^2$ + $\{ln[r_{comp}(\lambda=1.615 \,\mu\text{m}, \,\tau_m, D_{en}; \mu, \mu_0, \phi) - ln[r_{meas}(\lambda=1.615 \,\mu\text{m}, \,\mu, \mu_0, \phi)]\}^2$

Figure 16. Flow chart for cirrus solar retrieval program.

Determine combination of t<sub>m</sub>, Den that minimizes residual

+  $\{ln[r_{comp}(\lambda=2.132\mu m, \tau_m, D_{ep}; \mu, \mu_0, \phi) - ln[r_{meas}(\lambda=2.132\mu m, \mu, \mu_0, \phi)]\}^2$ 

### Daytime IR cirrus cloud retrieval program

Figure 17 shows the flowchart for the daytime IR cirrus cloud retrieval program. During the daytime, the 3.7 µm radiance contains both thermal emission and solar reflection. To apply the preceding retrieval program, we have developed an efficient but accurate scheme to remove the solar reflection part in the 3.7 µm radiance. This scheme is based on the correlations between 0.63 µm and 3.7 µm solar reflectances, which are obtained from the radiative transfer programs developed by Takano and Liou (1989). These programs take into account the scattering and absorption properties of hexagonal ice crystals for a range of mean effective ice crystal sizes and prescribed sun-satellite geometric parameters. The effects of the possible sources of errors on the retrieval results have been carefully analyzed in synthetic studies. Overall, the maximum error in the 3.7 µm solar component that can occur in the removal scheme is less than 10 percent (Rao et al. 1995). Validations of the retrieved cirrus optical depth and ice crystal size have been carried out using balloon-borne replicator and sounding data obtained during FIRE-II IFO. The balloon-borne replicator data can provide a nearly continuously vertical record of ice crystal size distributions in a Lagrangian sense. For validation purposes, an analytical method has also been developed to derive the optical depth and mean effective ice crystal size from the replicator data, taking into account the effects of different shapes and sizes (Ou et al., 1995, 1998). Figure 18 shows the flow chart for daytime IR cirrus cloud retrieval program.

# C3 Flow Chart for Cirrus Daytime IR Retrieval Algorithm Input VIIRS Data for Cloudy Pixels

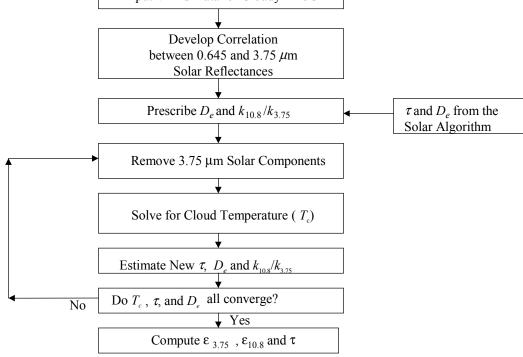


Figure 17. Flowchart for the daytime IR cirrus cloud retrieval program.

#### 3.3.2.2 Retrieval of Water Cloud Parameters

Once the cirrus cloud pixels are identified, two retrieval programs are applied to nighttime and daytime data as shown in Figure 18. For the daytime retrieval, the solar cloud retrieval algorithm (W1) using the VIIRS 0.645, 1.6, and 2.13  $\mu$ m radiance is employed to infer water cloud optical depth and mean effective size. For the nighttime retrieval, the IR cloud retrieval program (W2) using the VIIRS 3.75 and 10.8  $\mu$ m radiance is employed to infer the water cloud temperature, mean effective size, and visible optical depth.

### Solar water cloud retrieval scheme

A generalized schematic description of the solar water cloud retrieval algorithm is given as follows. Figure 19 shows an algorithm for retrieving the optical depth and mean effective size of cirrus clouds from comparisons of measured reflection functions with entries in the library. We first prescribe the sun-sensor geometric parameters, including the solar zenith angle, the sensor viewing zenith angle, and the relative azimuthal angle for each pixel. In application of this scheme to real-time satellite data, the surface albedo can be obtained from the histogram of clear reflectances. In the algorithm

SBRS Document #: Y2393 Raytheon

sensitivity studies, the surface albedo is determined based on climatological data. We then select values for the mean water droplet radius and optical depths.

# Water Cloud Retrieval Program

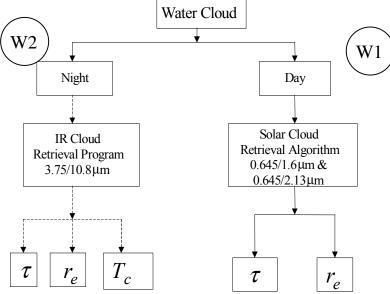


Figure 18. Flow chart for the water cloud retrieval program.

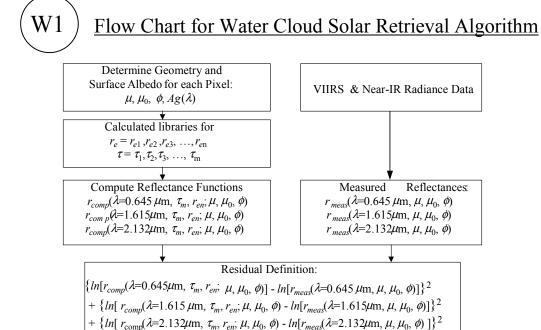


Figure 19. Flow chart for the water cloud solar retrieval program.

Determine combination of  $\tau_m$ ,  $D_{en}$  that minimizes residual

Subsequently, we construct radiance tables for each combination of water droplet effective radius and optical depth, and for each VIIRS channel used for cloud retrieval (0.645, 1.6, and section 3.3.1.3. A series of numerical iteration procedures are set up to search for the simulated reflectances that best match the measured reflectances. In the algorithm sensitivity studies, the "measured" reflectances are precomputed using the LBLE. The numerical scheme is basically composed of minimizing the sum of the square of the "residual" for each channel. The "residual" is defined as the difference between the logarithm of simulated reflectance and the logarithm of the measured reflectance.

### Water cloud IR retrieval program

The retrieval program for inferring water cloud temperature, effective particle size, and optical depth from the upwelling radiance of VIIRS cloud retrieval channels is similar to the cirrus cloud IR retrieval program. It also follows the principles of the dual-IR-channel technique presented in Liou *et al.* (1990). The VIIRS 3.75  $\mu$ m (Channel 1) and 10.8  $\mu$ m (Channel 2) radiance have been selected for the development. A major advantage of using these two channels for cirrus retrievals is that the presence of water vapor almost has no effect on the radiance of these window bands.

From the theory of radiative transfer, we may express the upwelling radiance at TOA for the 3.75 (Channel 1) and 10.8 µm (Channel 2) channels over a water cloud atmosphere as

$$R_i = R_{wi} + T_i R'_i i = 1, 2,$$
 (63)

where  $R_{wi}$  and  $T_i$  are the emitted radiance and the transmittance due to atmospheric water vapor and other gaseous species above the water cloud top, respectively; and  $R'_i$  is the atmospheric-corrected upwelling radiance at the cloud top. The second term in Eq. (63) represents the portion of the TOA radiance attributed to the transmitted upwelling radiance at the cloud top. A version of the radiative transfer program developed by Kratz (1995) has been used to compute the two atmospheric parameters. It is then possible to derive  $R'_i$  from Eq. (63).

We then further express  $R'_i$  in terms of the water cloud mean temperature  $T_c$  and emissivities  $\mathcal{E}_{1,2}$  as follows:

$$R'_{i} = (1 - \varepsilon_{i}) R'_{ai} + \varepsilon_{i} B_{i}(T_{c}) \quad i = 1, 2$$

$$(64)$$

where  $R'_{ai}$  denote the upwelling radiance reaching the cloud base for the two spectral bands and  $B_i(T_c)$  are the respective Planck functions at  $T_c$ . The first term on the right-hand side of Equation 64 represents the contribution of the transmitted radiance from below the cloud. The second term denotes the emission contribution from the cloud itself.

In order to solve Equation 64 for  $T_c$  and  $\varepsilon_i$  numerically, we need to correlate  $\varepsilon_1$  and  $\varepsilon_2$ , and express  $B_1(T_c)$  in terms of  $B_2(T_c)$ . The clear radiance  $R'_{a1,2}$  must also be known. First we compute the Planck functions  $B_1(T_c)$  and  $B_2(T_c)$ , taking into account the filter functions of both channels. A look-up table for both  $B_1(T_c)$  and  $B_2(T_c)$  in 0.1 K intervals is constructed by using a range of  $T_c$  from 150 to 300 K. Values in the look-up table are then fitted to a third-degree polynomial based on a least-square regression technique:

$$B_1(T_c) = \sum_{n=0}^{3} a_n [B_2(T_c)]^n = f[B_2(T_c)]$$
(65)

where the coefficients are  $a_0 = 2.6327 \times 10^{-4}$ ,  $a_1 = -1.063 \times 10^{-4}$ ,  $a_2 = 8.2976 \times 10^{-6}$ , and  $a_3 = 3.7311 \times 10^{-7}$ . Errors in the fitted polynomial are less than 1 percent.

Second, we investigate the relationship between the emissivities of the two channels. From radiative transfer calculations, and following the approach proposed by Liou *et al.* (1990), we may parameterize cirrus emissivities at the 3.75 and 10.8  $\mu$ m wavelengths in terms of visible optical depths  $\tau$  in the form:

$$\boldsymbol{\varepsilon}_{i} = \boldsymbol{\alpha}_{i}[1 - \exp(-k_{i}\boldsymbol{\tau})] \qquad \qquad i=1,2$$
(66)

The exponential term represents the effective transmissivity. The parameters  $k_1$  and  $k_2$  represent the effective extinction coefficients for the two channels, accounting for multiple scattering within cirrus clouds and for the difference between visible and IR extinction coefficients. Both  $k_1$  and  $k_2$  are smaller than 1 because the effect of multiple scattering increases transmission. Thus the products  $k_i\tau$  may be considered as effective optical depth that would yield the same emissivity values for the non-scattering conditions at the 3.75 and 10.8  $\mu$ m wavelengths. A scaling factor  $\alpha_i$  has been added to Eq (66), which accounts for the scattering effects. For the 10.8  $\mu$ m band,  $\alpha_2$  is nearly 1 due to very small scattering effects for that band. For the 3.75  $\mu$ m,  $\alpha_2$  is a function of  $r_e$ , and its value for a selected  $r_e$  can be obtained as the value of  $\varepsilon_i$  at very large  $\tau$  (~64). Using the  $\alpha_l$  values for the nine reference  $r_e$ , we obtain a polynomial function of  $\alpha_l$  in terms of  $r_e^{-1}$ , i. e.

$$\alpha_{1}(r_{e}) = \sum_{r=0}^{3} m_{n} r_{e}^{-n} . \tag{67}$$

By eliminating  $\tau$  from Equation 66, we obtain

$$(1 - \boldsymbol{\varepsilon}_{1} / \boldsymbol{\alpha}_{1})^{\frac{1}{k_{1}}} = (1 - \boldsymbol{\varepsilon}_{2})^{\frac{1}{k_{2}}} \tag{68}$$

Equation 68 correlates  $\varepsilon_1$  and  $\varepsilon_2$  directly. A further combination of Equations 64 and 66 leads to:

$$[1 - (\mathbf{R}_{a_1}^{\prime} - \mathbf{R}_{1}^{\prime}) / \boldsymbol{\alpha}_{1} (\mathbf{R}_{a_1}^{\prime} - \boldsymbol{B}_{1} (\boldsymbol{T}_{c}))]^{k_1} = [(\mathbf{R}_{2}^{\prime} - \boldsymbol{B}_{2} (\boldsymbol{T}_{c})) / \mathbf{R}_{a_2}^{\prime} - \boldsymbol{B}_{2} (\boldsymbol{T}_{c}))]^{k_2}$$
(69)

Substitution of Equation 64 into Equation 67 results in a nonlinear algebraic equation, with  $B_1(T_c)$  as the only unknown:

$$[(R'_2 - B_2(T_c))/(R'_{a2} - B_2(T_c))] - [1 - (R'_{a1} - R'_1)/(R'_{a1} - f[B_2(T_c)])]^{k_2/k_1} = 0$$
(70)

We investigate the dependence of  $k_2/k_1$  on the particle size distribution and optical depth based on radiative transfer calculations by using nine water droplet size distributions given in Section 3.4. A reasonable range of optical depth (0.125-64) for these size distributions has been used. As an initial approximation, we may take  $k_1$  and  $k_2$  as independent of the optical depth. However, for a more accurate retrieval of water cloud parameters, we need to consider the dependence of  $k_2/k_1$  on the optical depth. Generally,  $k_2/k_1$  decreases as  $r_e$  increases. For a small  $r_e$  (~2 µm),  $k_2/k_1$  is close to 2.5. This is primarily because the single-scattering albedo is larger for the 3.75 µm wavelength (~0.8) than that of the 10.8 µm wavelength (~0.4), which implies that more scattering is associated with the former wavelength. For  $r_e > 16$  µm,  $k_2/k_1$  approaches 1 for the following reasons. First, the extinction coefficients are approximately the same for the two wavelengths because of large-size parameters in which the geometric optics limit is valid. Second, the single-scattering albedos are also approximately the same for these wavelengths because substantial absorption occurs within large ice crystals. This implies that only the diffracted and externally reflected light contributes to the scattering processes. For obtaining the initial guesses,  $k_2/k_1$  can be expressed in terms of  $1/r_e$  in the form:

$$\mathbf{k}_{2} / \mathbf{k}_{1} = \sum_{n=0}^{2} \mathbf{b}_{n}^{*} \mathbf{r}_{e}^{-n}$$
 (71)

where the coefficients  $b'_n$  are determined from a second-order least square method using the radiance tables. In the retrieval scheme, however,  $b'_n$  are further expressed as third-order polynomial functions of  $\tau$ :

$$\boldsymbol{b}_{n}'(\boldsymbol{\tau}) = \sum_{n=0}^{3} \boldsymbol{d}_{n}' \boldsymbol{\tau}^{n} \tag{72}$$

where the coefficients  $d'_n$  are obtained from a third-order least square method.

A closure relationship for  $r_e$ ,  $\tau$ , and cloud liquid water content (*LWC*) is needed. From dimensional analysis:

$$r_e \propto (LWC)^{\frac{1}{3}},$$
 (73)

$$LWC \propto r_{e} \tau . \tag{74}$$

Combining Eqs (73) and (74), we obtain

$$r_{e} \propto \sqrt{\tau}$$
 (75)

Thus Equations 64 (2 equations), 65, 66 (2 equations), 70, and 75 form a complete set of seven governing equations for the solution of seven unknowns:  $\varepsilon_{1,2}$ ,  $B_{1,2}(T_c)$ ,  $T_c$ ,  $\tau$ , and  $r_e$ . Figure 20 shows a schematic description of the solution procedures. Using the VIIRS 3.75 and 10.8 µm channel data, we first obtain an estimate of mean water cloud temperature from an optimization scheme (Appendix D). Subsequently, an initial guess of  $\varepsilon_i$ ,  $\tau$ ,  $k_2/k_1$  and the proportional factor c (Equation 56) can be computed from the following equations:

$$\hat{\boldsymbol{\varepsilon}}_{2} = \frac{\boldsymbol{R}_{a2}^{\prime} - \boldsymbol{R}_{2}^{\prime}}{\boldsymbol{R}_{a2}^{\prime} - \boldsymbol{B}_{2}(\hat{\boldsymbol{T}}_{c})} \tag{76}$$

$$\hat{\boldsymbol{\tau}} = -\ln(1 - \hat{\boldsymbol{\varepsilon}}_2) / \hat{\boldsymbol{k}}_2 \tag{77}$$

$$\hat{k}_{2}/\hat{k}_{1} = \frac{\ln[R'_{2} - B_{2}(\hat{T}_{c})] - \ln[R'_{2} - B_{2}(\hat{T}_{c})]}{\ln[R'_{2} - B_{1}(\hat{T}_{c}) - (R'_{2} - R'_{1})/\alpha_{1}] - \ln[R_{2} - B_{1}(\hat{T}_{c})]}$$
(78)

where Equations 76 through 78 are the inverse form of Equations 64, 66, and 69. For the above initial guesses, we prescribe  $k_2$  as constant, although it is a function of effective particle size. Utilizing the estimated value of  $k_2/k_1$ , we proceed to solve Equation 69 for  $T_c$  by using a modified Newton's iteration scheme, which is efficient for the solution of nonlinear algebraic equations. Subsequently, we use Equations 64, 66, 70, and 75 to obtain new estimates of  $\varepsilon_i$ ,  $\tau$ ,  $k_2/k_1$ , and  $r_e$ . Then, we put the new estimate of  $T_c$ ,  $\tau$ , and  $r_e$  to convergence tests. If the value of the new estimate of these parameters is less than that of the previous iteration, then the new estimate of  $T_c$ ,  $\tau$ , and  $T_e$  is the solution. Otherwise, we go back to solve Equation 69 using the new estimate of  $k_2/k_1$ , and repeat the subsequent processes.

## Flow Chart for Cirrus Nighttime IR Retrieval Algorithm

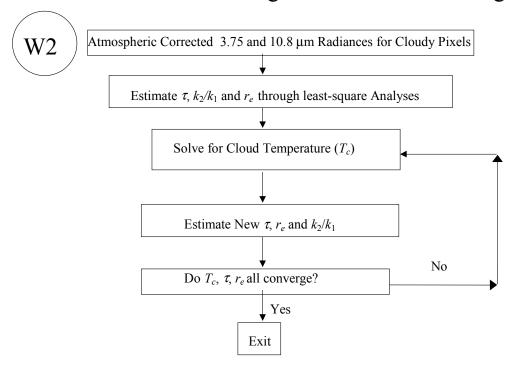


Figure 20. Flow chart for nighttime IR water cloud retrieval algorithm.

### 3.3.3 Archived Algorithm Output

For cirrus clouds, two retrieval algorithms will be used. The major products from the IR cloud retrieval program will include optical depth, mean effective size, and cloud temperature. The major products from the solar cloud retrieval program will be optical depth and mean effective size. The major products from the daytime IR cloud retrieval program will be the optical depth, mean effective size and cloud temperature.

For water clouds two retrieval algorithms will be used. The major products from the solar cloud retrieval program will be optical depth and mean effective size. The major products from the IR cloud retrieval program will be optical depth, mean effective size, and cloud temperature.

### 3.3.4 Variance and Uncertainty Estimates

The retrieval errors can be caused by instrumental characteristics including noise and calibration errors. Algorithm sensitivity studies on the EDR uncertainty due to instrumental noises have been performed by adding random noises to the simulated radiances according to the sensor noise model. Results of these studies are presented in sections 3.4 and 3.5. A summary of these studies is given as follows.

Algorithm sensitivity studies have been carried out separately for cirrus clouds using solar and IR algorithms and for water clouds using solar algorithms. These studies are designed to investigate whether the algorithms meet the threshold and objective requirements specified in VIIRS SRD (Appendix A), given reasonable assumptions about the uncertainty of input radiance. The key requirements for the retrieved optical depth and effective particle size are the measurement range, measurement accuracy, and precision errors. The other requirements are less relevant than these. According to the specification of the SRD, for optical depth, the threshold requirement for the measurement range is between 0 and 10. The threshold requirement for measurement accuracy is the greater of 10 percent or 0.05 optical depth; and the objective requirement for measurement accuracy is the greater of 5 percent or 0.025. Also, the threshold requirement for precision error is the greater of 5 percent or 0.025 optical depth, and the objective requirement is the greater of 2 percent or 0.02 optical depth. For effective particle size, the threshold requirement for the measurement range is between 0 and 50 µm. The threshold requirement for the measurement accuracy is the greater of 10 percent or 4 µm, and the objective requirement is the greater of 5 percent or 2 µm. The threshold requirement for the measurement precision is the greater of 5 percent or 2 µm, while the objective requirement is 2 percent. The results of algorithm sensitivity studies that these threshold and objective values for the measurement range, accuracy and precision errors are met.

A few values of the measurement requirements above are to be determined by the contractor (TBD). For optical depth, we suggest the TBD objective value for the measurement range is between 0.1 and 64, and between 0.1 and 12 for the retrieval of daytime and nighttime water clouds, respectively; and between 0.1 and 10 for the retrieval of both daytime and nighttime cirrus clouds. The lower limit of the range should be 0.1 instead of 0 for both threshold and objective. The originally set lower bound 0 corresponds to clear, while 0.1 represents an adequate lower limit based on research results (e.g., Ou et al., 1995). Moreover, the optical depths of water clouds are covered in most of this range. Established research-grade solar algorithms, e.g. King et al. (1996), address the range between 0.5 and 64 for water cloud retrieval. In addition, based on Ou et al. (1993, 1995) and Rao et al. (1995), the measurement range of the optical depth of both daytime and nighttime cirrus cloud is between 0.1 and 10. Nighttime retrievals for water clouds are more difficult to achieve, and the anticipated measurement range based on IR techniques is about 1 to 12.

We also suggest the objective value for the measurement precision of optical depth is the greater of 2 percent and 0.02. Based on Figure 10 of MODIS-ATBD (version 6.0), for water cloud optical depth larger than 5, the retrieval error of the solar algorithm is generally less than 2 percent. On the other hand, for ice clouds, Rao *et al.* (1995) show that for ice cloud optical depth larger than 1, the retrieval errors for optical depth are less than 2 percent, and for optical depth less than 1, the retrieval errors are less than 0.02 optical depth.

For cloud effective particle size, we suggest that the objective requirements for the measurement range are between 5 and 62.5  $\mu$ m for cirrus clouds, and between 2 and 32  $\mu$ m for water clouds. We specify the lower limit of the measurement range to be 1  $\mu$ m instead of 0  $\mu$ m, because particles with size smaller than 1  $\mu$ m are generally classified as aerosols. It has been demonstrated that the IR algorithm can address the ice crystal mean effective size ( $D_e$ ) range between 23 and 124 $\mu$ m (Ou *et al.*, 1993, 1995; Rao *et al.*, 1995). Because the definition of the effective particle size implies that this parameter corresponds to the dimension of radius, the above range corresponds to the range of effective particle size between 11.5 and 62  $\mu$ m. With available *in situ* contrail size distributions and associated single-scattering properties based on the unified light-scattering theories for ice crystals

(Liou et al., 1998), the lower limit of retrievable ice crystal effective particle size can be extended down to 5  $\mu$ m. On the other hand, established research-grade solar algorithms address the measurement range between 2 and 32  $\mu$ m for water clouds. Nighttime retrievals for water clouds are more difficult to achieve, and the anticipated measurement range based on the IR techniques is expected to be from 4 to 16  $\mu$ m.

### 3.4 ALGORITHM SENSITIVITY STUDIES ON RETRIEVALS OF OPTICAL THICKNESS

The algorithm sensitivity studies were carried out using the results generated by the radiative transfer model for seven separate scenarios. These scenarios form the basis for defining the input parameters required by the radiative transfer model. They are:

- (1) Cirrus cloud in US Standard Atmosphere, sensor pointing at nadir.
- (2) Cirrus cloud in US Standard Atmosphere, sensor pointing at off-nadir.
- (3) Cirrus cloud in US Standard Atmosphere, sensor pointing at edge-of-scan.
- (4) Cirrus cloud in Tropical Atmosphere, sensor pointing at nadir.
- (5) Cirrus cloud in Sub-arctic Atmosphere, sensor pointing at nadir.
- (6) Cirrus cloud in Desert Atmosphere, sensor pointing at nadir.
- (7) Water cloud in US Standard Atmosphere, sensor pointing at nadir.

For scenarios 1 through 4, synthetic retrievals for both daytime and nighttime conditions were carried out for combinations of input cirrus optical depths and effective particle sizes. For scenario 7, only synthetic retrievals for daytime conditions were carried out.

The LBLE described in Section 3.3.1.7 was used to generate simulated radiance. The input parameters for each scenario are given in Table 6. For all scenarios, the solar zenith angle is fixed at 32°, the nominal date is 1 May and the location is 40° North based on the Raytheon VIIRS orbital simulation. Parameters specified include the atmospheric profile, sun-sensor geometry, retrieval channel characteristics (including central wavelength, band-width and response function), cloud type, altitude and thickness, as well as surface albedo and emissivity. The radiative transfer calculations are performed for a wide range of selected visible optical depth and mean effective size. For cirrus clouds, the optical depths chosen are: 0.125, 0.25, 0.5, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, and 12. The mean effective sizes chosen are: 23.9, 30.4, 41.5, 71, and 93 μm. For water clouds the optical depths selected are: 0.125, 0.25, 0.5, 1, 2, 3, 4, 5, 6, 8, 12, 16, 24, 32, 48, and 64. The mean droplet radii selected are: 2, 3, 4, 6, 8, 12, 16, 24, and 30 μm. Many radiance values are thus produced, and they are further processed to create data sets that are appropriate for the parameter retrieval algorithms and algorithm sensitivity analyses.

This section addresses results of the algorithm sensitivity analysis for Cloud Optical Depth using the cirrus and water cloud IR and Solar retrieval algorithms, which use VIIRS bands at 3.75 and 10.8  $\mu$ m and bands at 0.645, 1.610, and 2.130  $\mu$ m, respectively. The results of these algorithm sensitivity analyses follow.

### 3.4.1 SNR Tests

The SNR tests address the impact of SNR on both pixel-level and image-level retrievals. The pixel-level retrievals are performed for a single pixel assuming various combinations of cloud optical depth and effective particle size. The imagery-level retrievals are performed for pixels aggregated (averaged) to VIIRS SRD horizontal cell size (HCS). The required HCS is a function of EDR parameter; threshold and objective values are stated. *In this report, only the pixel-level SNR tests are addressed.* The test results are presented separately for the daytime (solar algorithm) and nighttime (IR) algorithms. To create data sets to support the pixel level SNR test, noise based on the sensor noise model is added directly to the radiance contained in the radiance tables and then retrievals are performed using the noise-added data in the tables. To guarantee sample sizes are sufficient to support tests, noise is randomly added to each radiance value 32 times. In effect, 32 noise perturbed radiance tables are created. Retrievals are performed using each of the 32 noise-added tables and the no-noise table. The results of the retrieval process are then statistically analyzed to compute the metrics described in the SRD (accuracy, precision, etc.).

**Table 6. Input Parameters for Each Scenario** 

	SET#								
	USSCir-n1*	USSCir-i*	USSCir-eos	TCir-n	SACir-n	DesCir-n	USSWat-n		
Description	Baseline US Standard/ Cirrus Nadir		Baseline US Standard/ Cirrus	Baseline Tropical/ Cirrus Nadir	Baseline Subarctic/ Cirrus Nadir	Baseline Desert/ Cirrus Nadir	Baseline US Standard/ Water		
		Intermediate	Edge-of- Scan				Cloud		
Atmosphere									
US Standard (1)	X	X	X				X		
Tropical (3)				Х					
Sub-arctic					Х				
Desert						X			
Skin									
Temperature MODTRAN	X	X	X	X	X	X	X		
Default (uses sfc									
air temp.)									
Aerosol Model MODTRAN	X	X	Х	X	X		Х		
Default	^	^	^	<b>X</b>	^		Α		
Desert						Х			
Sensor									
Geometry									
Nadir	Χ			Χ	Χ	Χ	Χ		
Off-Nadir		Х			•				
Edge-of-Scan			X			4			
Cloud Height	9-10 km	9-10 km	9-10 km	14-15 km	9-10 km	9-10 km	1 -2 km		

### 3.4.1.1 Daytime Results

### Cirrus cloud in US Standard Atmosphere, sensor pointing at nadir

Figure 21 shows the two-dimensional correlations for the reflectance pairs:  $0.645 - 1.6 \,\mu m$  and  $0.645 - 2.1 \,\mu m$  for the selected six ice crystal size distributions and 14 optical depths. These diagrams clearly illustrate the information content of optical depth and mean effective size in the 0.645, 1.6 and 2.1  $\,\mu m$  reflectances. The 0.645 and 1.6  $\,\mu m$  reflectance correlation shows that if the mean effective size of an ice cloud is larger than about 24  $\,\mu m$  but less than about 124  $\,\mu m$ , an accurate retrieval of ice crystal mean effective size can be achieved. Based on a similar argument, if the optical depth of an ice cloud is larger than about 0.5 but less than 12, an accurate retrieval of optical depth can also be obtained. Also noted is that the 0.645  $\,\mu m$  reflectance mainly depends on the optical depth, while the 1.6 and 2.1  $\,\mu m$  reflectances are primarily functions of mean ice crystal size for optically thick ice

SBRS Document #: Y2393 Raytheon

clouds. Consequently, data for these three channels can be utilized to determine both optical depth and mean effective ice crystal size. The validity of such correlations has been examined using the MAS data obtained from the flight missions during FIRE-II IFO. Limited comparisons between MAS data points and computed results show that most of the data points fall inside the correlative mesh of optical depth vs. mean effective size.

Figure 22 show the accuracy and precision of retrieved optical depths as functions of optical depth. These results are based on retrievals of combinations of all possible optical depths and mean effective sizes using the display of the 0.645 and 1.6 µm reflectance correlation. Two straight lines denote threshold (greater of 10 percent or 0.05) and objective (greater of 5 percent or 0.025) values. The accuracy of retrieved optical depths meets both the threshold and objective requirements for most selected values of optical depth. The precision of retrieved optical depths meets both the threshold (greater of 5 percent or 0.025) and objective (greater of 2 percent or 0.02) requirements for most selected values of optical depth. For very small optical depths (e.g., < 0.5), both the accuracy and precision errors are close to threshold and objective requirements, but for larger optical depths, both the accuracy and precision errors are well below the threshold and objective curves and are close to zero. The extremely small values of accuracy and precision errors indicate that the solar algorithm is highly accurate and very stable in retrieving optical depths in this case.

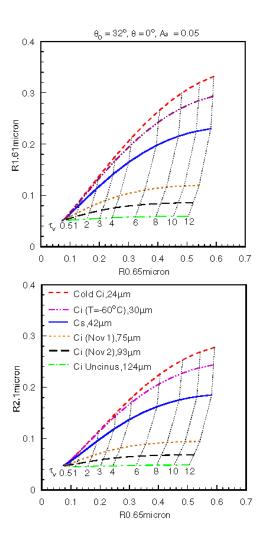
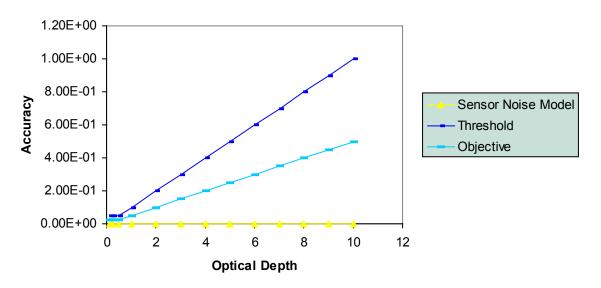


Figure 21. The correlation between the VIIRS 0.645 and 1.6  $\mu$ m reflectances and between the 0.645 and 2.13  $\mu$ m reflectances for cirrus clouds in US Standard Atmosphere with the sensor pointing at nadir.

# **Accuracy of Optical Depth (Daytime)**



# **Precision of Optical Depth (Daytime)**

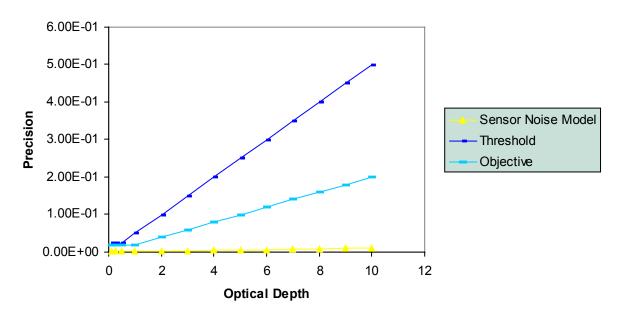


Figure 22. Accuracy and Precision of retrieved optical depths from the cirrus solar algorithm for US Standard Atmosphere, sensor at nadir, based on the  $0.645-1.61~\mu m$  correlation.

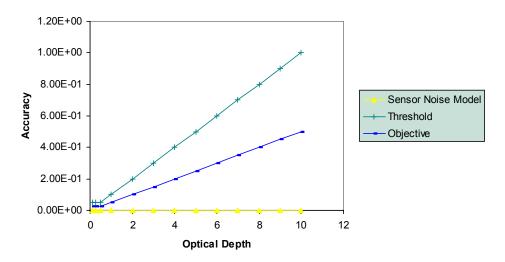
To examine the effects of adding the VIIRS 2.13 µm channel for the retrieval of cirrus cloud parameters, Figure 23 show the accuracy and precision of retrieved optical depths as functions of

optical depth. These results are based on retrievals of combination of all possible optical depths and mean effective sizes using the display of the 0.645 and 2.13  $\mu$ m reflectance correlation. Two straight lines denote threshold (greater of 10 percent or 0.05) and objective (greater of 5 percent or 0.025) values. The accuracy of retrieved optical depths meets both the threshold and objective requirements for most selected values of optical depth. The precision of retrieved optical depths meets both the threshold (greater of 5 percent or 0.025) and objective (greater of 2 percent or 0.02) requirements for most selected values of optical depth. For very small optical depths (e.g., < 0.5), both the accuracy and precision errors are close to threshold and objective requirements. For larger optical depths, both the accuracy and precision errors are close to zero for a variety of surface conditions, different geometries, and different cloud types. Figure 23 is similar to Figure. 22. Thus we expect all retrievals using the 0.645-2.13  $\mu$ m correlations will produce similar results as those using the 0.645-1.6  $\mu$ m correlations. For the rest of the algorithm studies on the solar retrieval algorithm, we will only show results using the 0.645-1.6  $\mu$ m correlation.

# Cirrus cloud in US Standard Atmosphere, sensor at off-nadir direction

Figure 24 shows the two-dimensional correlations for the reflectance pairs:  $0.645 - 1.6 \,\mu m$  and  $0.645 - 2.13 \,\mu m$  for the selected ice crystal size distributions and optical depths. The sensor zenith angle is  $40^{\circ}$ , and the relative azimuthal angle is  $136^{\circ}$ . Like Figure 21, these diagrams illustrate the information content of optical depth and mean effective size in the 0.645, 1.6, and  $2.1 \,\mu m$  reflectances. However, both the 1.6, and  $2.1 \,\mu m$  surface albedos are changed from 0.05 in Figure 21 to 0.25 to simulate the land surface condition. Thus, the shape of the correlation mesh in Figure 24 is different than that in Figure 21. This change of the shape of the correlation mesh does not affect the retrieval accuracy and precision. Display of the 0.645 and  $1.6 \,\mu m$  reflectances correlation shows that if the mean effective size of an ice cloud is larger than about  $24 \,\mu m$  but less than about  $124 \,\mu m$ , an accurate retrieval of ice crystal mean effective size can be achieved. Based on a similar argument, if the optical depth of an ice cloud is larger than about  $0.5 \,\mu m$  but less than  $12 \,\mu m$  accurate retrieval of optical depth can also be obtained. Also noted is that the  $0.645 \,\mu m$  reflectance mainly depends on the optical depth, while the  $1.6 \,\mu m$  reflectances are primarily functions of mean ice crystal size for optically thick ice clouds. Consequently, data for these three channels can be utilized to determine both optical depth and mean effective ice crystal size.

#### **Accuracy of Cloud Optical Depth**



# **Precision of Cirrus Cloud Optical Depth**

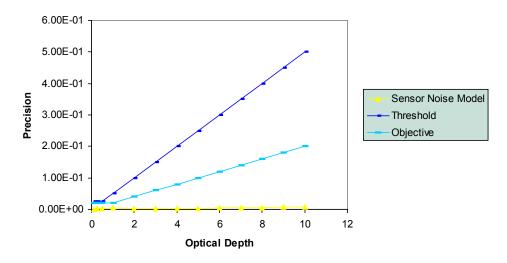


Figure 23. Accuracy and Precision of retrieved optical depths from the cirrus solar algorithm for US Standard Atmosphere, sensor at nadir, based on the  $0.645-2.13~\mu m$  correlation.

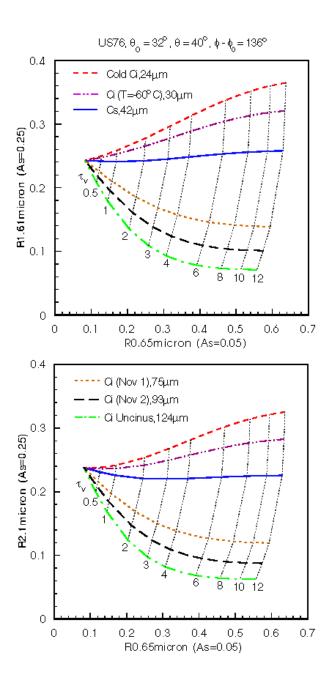


Figure 24. The 0.645-1.6 μm and 0.645-2.13 μm reflectance correlations for cirrus cloud in US Standard Atmosphere, sensor at off-nadir.

Figure 25 shows the accuracy and precision of retrieved optical depths as functions of optical depth. These results are based on retrievals of combinations of all possible optical depths and mean effective sizes. Two straight lines denote threshold (greater of 10 percent or 0.05) and objective (greater of 5 percent or 0.025) values. The accuracy of retrieved optical depths meets both the threshold and objective requirements for most selected values of optical depth. The precision of retrieved optical depths meets both the threshold (greater of 5 percent or 0.025) and objective (greater of 2 percent or 0.02) requirements for most selected values of optical depth. For very small optical depths (e.g., <

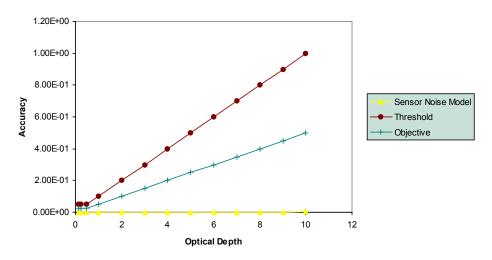
0.5), both the accuracy and precision errors are close to threshold and objective requirements, but for larger optical depths, both the accuracy and precision errors are well below the threshold and objective curves. The small values of accuracy and precision errors again indicate that the solar algorithm is highly accurate and very stable in retrieving optical depths for off-nadir sensor position.

## Cirrus cloud in US Standard Atmosphere, sensor at edge-of-scan direction

Figure 26 shows the two-dimensional correlations for the reflectance pairs:  $0.645 - 1.6 \,\mu m$  and  $0.645 - 2.1 \,\mu m$  for the selected ice crystal size distributions and optical depths. The sensor is assumed to be pointing at edge-of-scan. Like Figure 21, these diagrams illustrate the information content of optical depth and mean effective size in the 0.645, 1.6, and  $2.1 \,\mu m$  reflectances. However, both the 1.6, and  $2.1 \,\mu m$  surface albedos are changed from 0.05 in Figure 21 to 0.25 to simulate the land surface condition. Thus, the shape of the correlation mesh in Figure 26 is different than that in Figure 21. This change of the shape of the correlation mesh does not affect the retrieval accuracy and precision. The 0.645 and  $1.6 \,\mu m$  reflectance correlation shows that if the mean effective size of an ice cloud is larger than about  $24 \,\mu m$  but less than about  $124 \,\mu m$ , an accurate retrieval of ice crystal mean effective size can be achieved. Likewise, if the optical depth of an ice cloud is larger than about  $0.5 \,\mu m$  reflectance mainly depends on the optical depth, while the  $1.6 \,\mu m$  and  $2.1 \,\mu m$  reflectances are primarily functions of mean ice crystal size for optically thick ice clouds. Consequently, data for these three channels can be utilized to determine both optical depth and mean effective ice crystal size.

Figure 27 shows the accuracy and precision of retrieved optical depths as functions of optical depth. These results are based on retrievals of combination of all possible optical depths and mean effective sizes. Two straight lines denote threshold (greater of 10 percent or 0.05) and objective (greater of 5 percent or 0.025) values. The accuracy of retrieved optical depths meets both the threshold and objective requirements for most selected values of optical depth. The precision of retrieved optical depths meets both the threshold (greater of 5 percent or 0.025) and objective (greater of 2 percent or 0.02) requirements for most selected values of optical depth. For very small optical depths (e.g., < 0.5), both the accuracy and precision errors are close to threshold and objective requirements, but for larger optical depths, both the accuracy and precision errors are well below the threshold and objective curves. The small values of accuracy and precision errors again indicate that the solar algorithm is highly accurate and very stable in retrieving optical depths in the edge-of-scan case.

#### **Accuracy of Cirrus Cloud Optical Depth**



#### **Precision of Optical Depth**

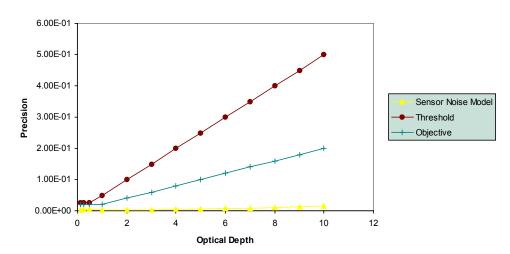


Figure 25. Accuracy and Precision of retrieved optical depths from the cirrus solar algorithm for US Standard Atmosphere, sensor at off-nadir.

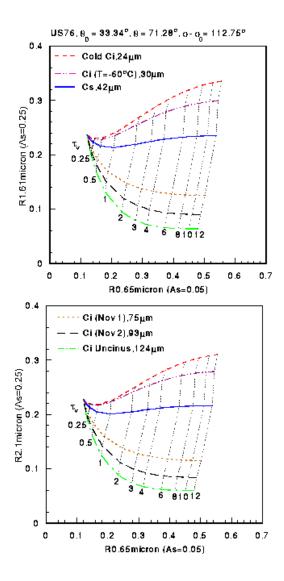
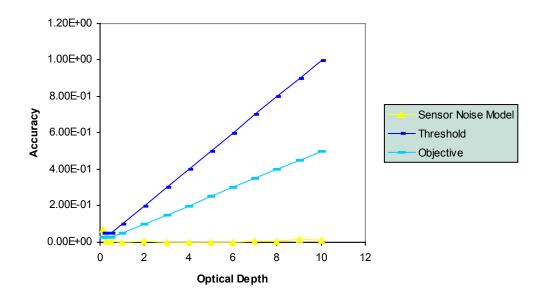


Figure 26. Display of the 0.645-1.6  $\mu m$  and 0.645-2.13  $\mu m$  reflectance correlations for cirrus cloud in US Standard Atmosphere, sensor at edge-of-scan.

Accuracy - US Standard Atmosphere, ice cloud optical depth, solar, edge-of-scan



Precision - Optical Depth, US Standard Atmosphere, Ice cloud , solar, edge of scan

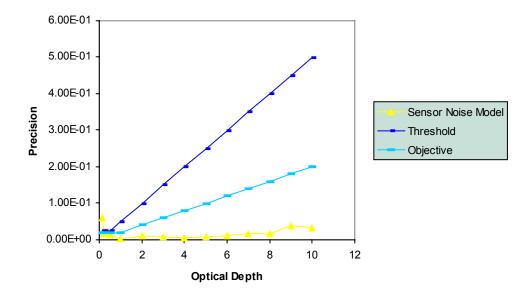


Figure 27. Accuracy and Precision of retrieved optical depths from the solar algorithm for cirrus cloud in US Standard Atmosphere with sensor pointing at edge-of-scan.

#### Cirrus cloud in tropical atmosphere with sensor at nadir direction

Figure 28 shows the two-dimensional correlations for the reflectance pairs:  $0.645 - 1.6 \,\mu m$  and  $0.645 - 2.1 \,\mu m$  for the selected ice crystal size distributions and optical depths. The sensor is assumed to be pointing at nadir. Like Figure 21, these diagrams illustrate the information content of optical depth and mean effective size in the 0.645, 1.6, and  $2.1 \,\mu m$  reflectances. However, both the 1.6, and  $2.1 \,\mu m$  surface albedos are changed from 0.05 in Figure 21 to 0.25 to simulate the land surface condition. Thus, the shape of the correlation mesh in Figure 28 is different than that in Figure 21. This change of the shape of the correlation mesh does not affect the retrieval accuracy and precision. Display of the 0.645 and  $1.6 \,\mu m$  reflectance correlation shows that if the mean effective size of an ice cloud is larger than about 24  $\,\mu m$  but less than about  $124 \,\mu m$ , an accurate retrieval of ice crystal mean effective size can be achieved. Likewise, if the optical depth of an ice cloud is larger than about 0.5 but less than 12, an accurate retrieval of optical depth can also be obtained. Also noted is that the  $0.645 \,\mu m$  reflectance mainly depends on the optical depth, while the 1.6 and  $2.1 \,\mu m$  reflectances are primarily functions of mean ice crystal size for optically thick ice clouds. Consequently, data for these three channels can be utilized to determine both optical depth and mean effective ice crystal size.

Figures 29 shows the accuracy and precision of retrieved optical depths as functions of optical depth. These results are based on retrievals of combination of all possible optical depths and mean effective sizes. Two straight lines denote threshold (greater of 10 percent or 0.05) and objective (greater of 5 percent or 0.025) values. The accuracy of retrieved optical depths meets both the threshold and objective requirements for most selected values of optical depth. The precision of retrieved optical depths meets both the threshold (greater of 5 percent or 0.025) and objective (greater of 2 percent or 0.02) requirements for most selected values of optical depth. For very small optical depths (e.g., < 0.5), both the accuracy and precision errors are close to threshold and objective requirements, but for larger optical depths, both the accuracy and precision errors are well below the threshold and objective curves. The small values of accuracy and precision errors again indicate that the solar algorithm is highly accurate and very stable in retrieving optical depths.

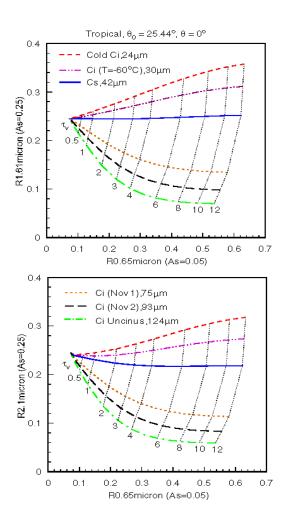
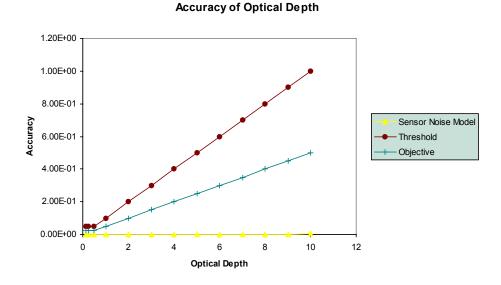


Figure 28. The 0.645-1.61  $\mu m$  and 0.645-2.13  $\mu m$  reflectance correlations for cirrus clouds in the Tropical Atmosphere, sensor at nadir.



# 

Figure 29. Accuracy and Precision of retrieved optical depths from the solar algorithm for cirrus cloud in Tropical Atmosphere with sensor pointing at nadir.

Optical Depth

# Cirrus cloud in subarctic atmosphere with sensor at nadir direction

Figure 30 shows the two-dimensional correlation for the reflectance pair:  $1.2-1.6~\mu m$  for the selected ice crystal size distributions and optical depths. The sensor is assumed to be pointing at nadir. Like Figure 21, these diagrams illustrate the information content of optical depth and mean effective size in the 1.2 and 1.6 $\mu m$  reflectances. However, the 1.2 and 1.6 $\mu m$  surface albedos are changed from 0.05 in Figure 21 to 0.164 and 0.0 respectively to simulate the ice/snow surface condition. Thus, the shape of the correlation mesh in Figure 30 is different than that in Figure 21. This change of the

70

shape of the correlation mesh does not affect the retrieval accuracy and precision. Display of the 0.645 and 1.6  $\mu$ m reflectance correlation shows that if the mean effective size of an ice cloud is larger than about 24  $\mu$ m but less than about 124  $\mu$ m, an accurate retrieval of ice crystal mean effective size can be achieved. Likewise, if the optical depth of an ice cloud is larger than about 0.5 but less than 12, an accurate retrieval of optical depth can also be obtained. Also noted is that the 0.645  $\mu$ m reflectance mainly depends on the optical depth, while the 1.6 and 2.1  $\mu$ m reflectances are primarily functions of mean ice crystal size for optically thick ice clouds. Consequently, data for these three channels can be utilized to determine both optical depth and mean effective ice crystal size.

Figures 31 shows the accuracy and precision of retrieved optical depths as functions of optical depth. These results are based on retrievals of combination of all possible optical depths and mean effective sizes. Two straight lines denote threshold (greater of 10 percent or 0.05) and objective (greater of 5 percent or 0.025) values. The accuracy of retrieved optical depths meets both the threshold and objective requirements for most selected values of optical depth. The precision of retrieved optical depths meets both the threshold (greater of 5 percent or 0.025) and objective (greater of 2 percent or 0.02) requirements for most selected values of optical depth. For very small optical depths (e.g., < 0.5), both the accuracy and precision errors are close to threshold and objective requirements, but for larger optical depths, both the accuracy and precision errors are well below the threshold and objective curves. The small values of accuracy and precision errors again indicate that the solar algorithm is highly accurate and very stable in retrieving optical depths.

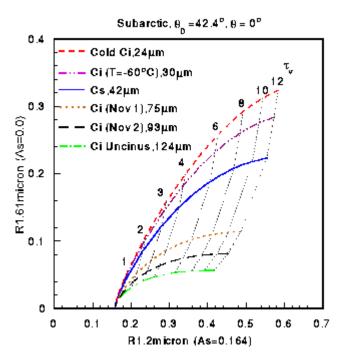
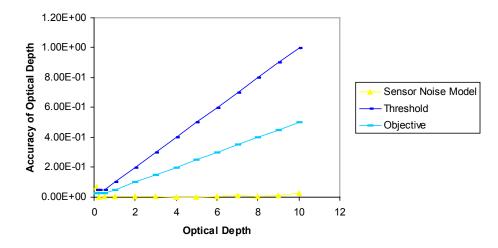


Figure 30. The 0.645-1.6 µm reflectance correlations for cirrus cloud in Subarctic Atmosphere, sensor at nadir.

#### Accuracy of Optical Depth (Subarctic, Solar Ice Cloud, Nadir)



#### Precision of Optical Depth (Subarctic, Solar Ice cloud, Nadir)

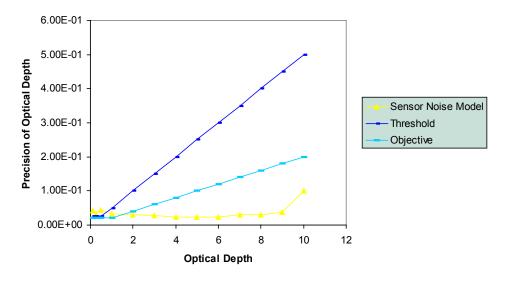


Figure 31. Accuracy and Precision of retrieved optical depths from the solar algorithm for cirrus cloud in Subarctic Atmosphere with sensor pointing at nadir.

# Cirrus cloud in desert atmosphere with sensor at nadir direction

Figure 32 shows the two-dimensional correlations for the reflectance pairs:  $0.645 - 1.6 \mu m$  for the selected ice crystal size distributions and optical depths. The sensor is assumed to be pointing at nadir.

Like Figure 21, these diagrams illustrate the information content of optical depth and mean effective size in the 0.645 and 1.6  $\mu$ m reflectances. However, both the 0.645 and 1.6  $\mu$ m surface albedos are changed from 0.05 in Figure 21 to 0.40 and 0.64, respectively, to simulate the land surface condition. Thus, the shape of the correlation mesh in Figure 32 is different than that in Figure 21. This change of the shape of the correlation mesh does not affect the retrieval accuracy and precision. Display of the 0.645 and 1.6  $\mu$ m reflectance correlation shows that if the mean effective size of an ice cloud is larger than about 24  $\mu$ m but less than about 124  $\mu$ m, an accurate retrieval of ice crystal mean effective size can be achieved. Likewise, if the optical depth of an ice cloud is larger than about 0.5 but less than 12, an accurate retrieval of optical depth can also be obtained. Also noted is that the 0.645  $\mu$ m reflectance mainly depends on the optical depth, while the 1.6  $\mu$ m reflectances are primarily functions of mean ice crystal size for optically thick ice clouds. Consequently, data for these three channels can be utilized to determine both optical depth and mean effective ice crystal size.

Figures 33 shows the accuracy and precision of retrieved optical depths as functions of optical depth. These results are based on retrievals of combination of all possible optical depths and mean effective sizes. Two straight lines denote threshold (greater of 10 percent or 0.05) and objective (greater of 5 percent or 0.025) values. The accuracy of retrieved optical depths meets both the threshold and objective requirements for most selected values of optical depth. The precision of retrieved optical depths meets the threshold (greater of 5 percent or 0.025) requirements for most selected values of optical depth. The small values of accuracy and precision errors again indicate that the solar algorithm is highly accurate and very stable in retrieving optical depths.

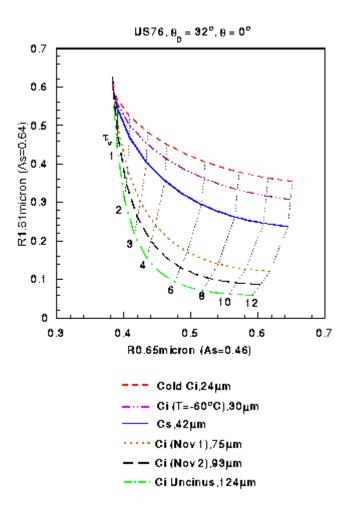
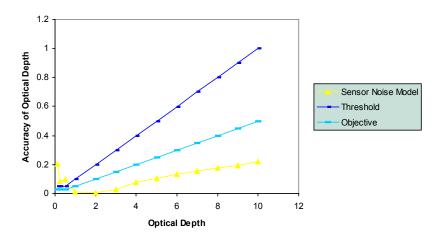


Figure 32. The 0.645-1.6 µm reflectance correlations for cirrus cloud in Desert Atmosphere, sensor at nadir.

# Accuracy of Optical Depth (Desert, US Standard Atm., Solar Ice Clouds, Nadir)



# Precision of Optical Depth (Desert, US Standard Atmosphere, Solar Ice Cloud, Nadir)

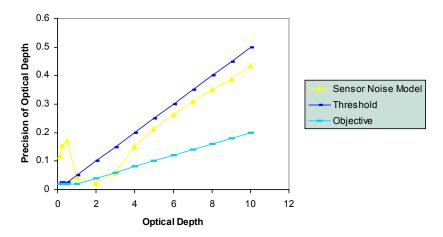


Figure 33. Accuracy and Precision of retrieved optical depths from the solar algorithm for cirrus cloud in Desert Atmosphere with sensor pointing at nadir.

#### Water cloud in US Standard Atmosphere, sensor pointing at nadir

Figure 34 shows the two-dimensional correlations for the reflectance pairs:  $0.645 - 1.6 \,\mu m$  and  $0.645 - 2.1 \,\mu m$  for the selected water droplet effective radius and optical depths. The sensor is assumed to be pointing at nadir. These diagrams illustrate the information content of optical depth and droplet effective radius in the 0.645, 1.6, and  $2.1 \,\mu m$  reflectances. Both the 1.6, and  $2.1 \,\mu m$  surface albedos are set at 0.05 as in Figure 21 to approximately simulate the ocean surface condition. Thus, the shape

of the correlation mesh in Figure 32 is similar to that in Figure 21. This similarity in the shape of the correlation mesh causes the retrieval accuracy and precision to be close to those for cirrus clouds in US Standard Atmosphere with sensor pointing at the nadir direction. Display of the 0.645 and  $1.6 \mu m$  reflectance correlation shows that if the droplet effective radius of a water cloud is larger than about 4  $\mu m$  but less than about 30  $\mu m$ , an accurate retrieval of water droplet effective radius can be achieved.

The curves of 2 and 3  $\mu$ m partially overlap those of 4 and 6  $\mu$ m. Therefore, if the droplet effective size is less than 4  $\mu$ m, the retrieved optical depths may be in error, because the retrieval scheme will take the data as from a cloud of 5 and 6  $\mu$ m. A correction scheme will be developed in the future to solve this problem. If the optical depth of an ice cloud is larger than about 0.5 but less than 12, an accurate retrieval of optical depth can also be obtained. It is noted that the 0.645  $\mu$ m reflectance mainly depends on the optical depth, while the 1.6 and 2.1  $\mu$ m reflectances are primarily functions of mean ice crystal size for optically thick ice clouds. Consequently, data for these three channels can be utilized to determine both optical depth and droplet effective radius.

Figure 35 show the accuracy and precision of retrieved optical depths as functions of optical depth. These results are based on retrievals of combinations of all possible optical depths and mean effective sizes. Two straight lines denote threshold (greater of 10 percent or 0.05) and objective (greater of 5 percent or 0.025) values. The accuracy of retrieved optical depths meets both the threshold and objective requirements for most selected values of optical depth. The precision of retrieved optical depths meets both the threshold (greater of 5 percent or 0.025) and objective (greater of 2 percent or 0.02) requirements for most selected values of optical depth. For very small optical depths (e.g., < 0.5), both the accuracy and precision errors are close to threshold and objective requirements, but for larger optical depths, both the accuracy and precision errors are well below the threshold and objective curves. The small values of accuracy and precision errors again indicate that the solar algorithm is highly accurate and very stable in retrieving optical depths for water clouds in US Standard Atmosphere with the sensor pointing at nadir.

76

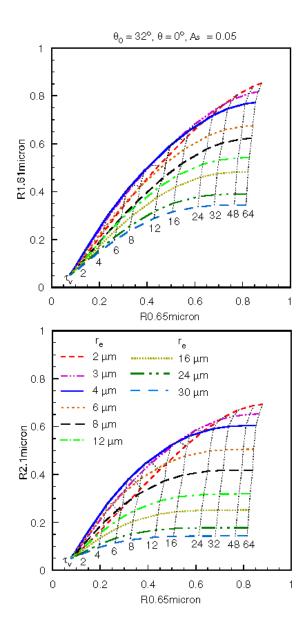
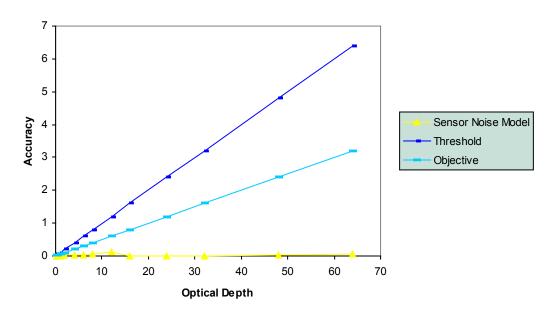


Figure 34. The 0.645-1.6  $\mu m$  and 0.645-2.13  $\mu m$  reflectance correlations for cirrus cloud in US Standard Atmosphere, sensor at nadir.

#### Accuracy of Optical Depth for Water Cloud at Nadir



#### Precision of Optical Depth for Water Cloud at nadir

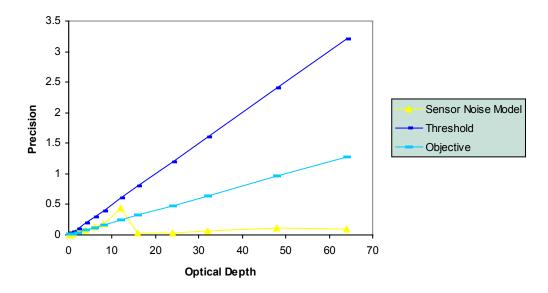


Figure 35. Accuracy and Precision of retrieved optical depths from the solar algorithm for water cloud in US Standard Atmosphere with sensor pointing at nadir.

### 3.4.1.2 Nighttime Results

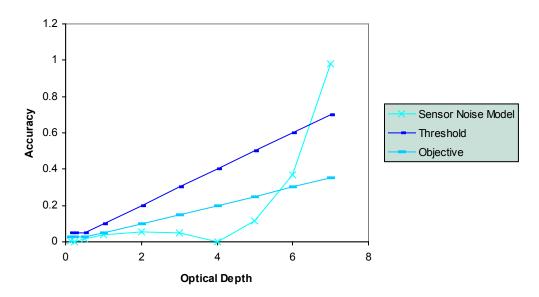
# Cirrus cloud in US Standard Atmosphere, sensor pointing at nadir

Figure 36 shows the accuracy and precision of retrieved optical depths from the nighttime IR algorithm as a function of optical depth. These results are based on retrievals of combinations of optical depths between 0.125 and 7 and of mean effective sizes between 20 and 100 µm. Two straight lines denote threshold (greater of 10 percent or 0.05) and objective (greater of 5 percent or 0.025) values. The accuracy of the retrieved optical depths meets the threshold requirement for optical depths up to 6. The accuracy does not meet the threshold for optical depths larger than 6, mainly because the exponential relationship between the IR emissivity and the optical depth. For optically thick clouds, the IR emissivity approaches one, so that a small error in the retrieved emissivity can lead to a large error in optical depth. This behavior is very different from the results of the retrieval of the cirrus cloud optical depth using the solar algorithm. The precision of retrieved optical depths meets the threshold (greater of 5 percent or 0.025) requirement for optical depths up to 6. Precision becomes less satisfactory for larger optical depths, because the sensitivity of emissivity toward optical depth is reduced due to the exponential relationship between the emissivity and optical depth.

## Cirrus coud in US Standard Atmosphere, sensor pointing at off-nadir

Figure 37 shows the accuracy and precision of retrieved optical depths from the nighttime IR algorithm as a function of optical depths. These results are based on retrievals of combinations of optical depths between 0.125 and 7 and of mean effective sizes between 20 and 100 µm. The straight lines denote threshold (greater of 10 percent or 0.05) and objective (greater of 5 percent or 0.025) values. The accuracy of the retrieved optical depths meets the threshold requirement for optical depths up to 7. The accuracy does not meet the threshold for optical depths larger than 7, mainly due to the exponential relationship between the IR emissivity and the optical depth. For optically thick clouds, the IR emissivity approaches one, so that a small error in the retrieved emissivity can lead to a large error in optical depth. This behavior is very different from the results of the retrieval of the cirrus cloud optical depth using the solar algorithm. The precision of retrieved optical depths meets the threshold (greater of 5 percent or 0.025) requirement for optical depths up to 5. Precision becomes less satisfactory for larger optical depths, because sensitivity of emissivity toward optical depth is reduced due to the exponential relationship between the emissivity and optical depth.

# **Accuracy of Optical Depth**



# Precision of Optical Depth for De = 41.5 um

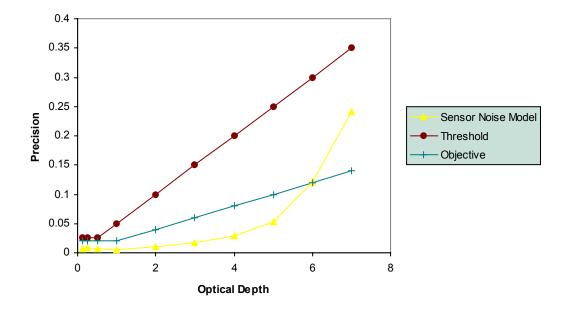
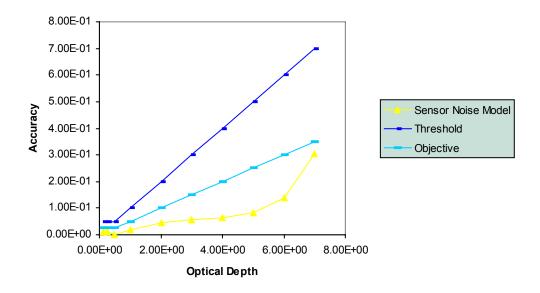


Figure 36. Accuracy and Precision of retrieved optical depths from the IR algorithm for cirrus cloud in US Standard Atmosphere with sensor pointing at nadir.

# Accuracy of Cirrus Cloud Optical Depth from Nighttime IR Algorithm for US Standard Atmosphere at theta=40 deg



# **Precision of Optical Depth**

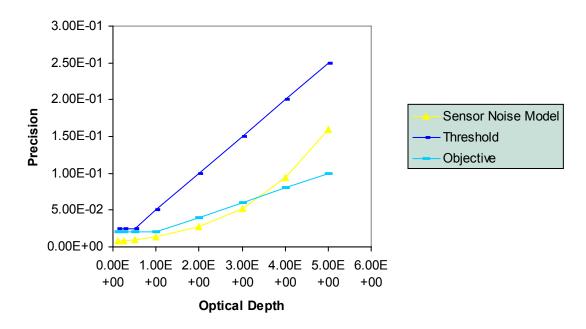
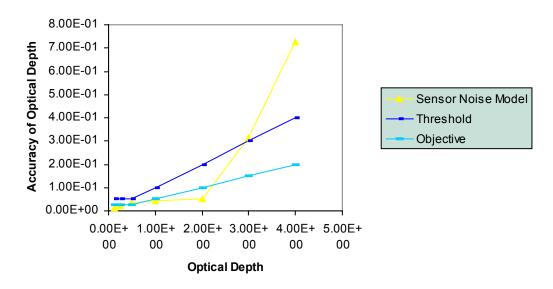


Figure 37. Accuracy and Precision of retrieved optical depths from the IR algorithm for cirrus cloud in US Standard Atmosphere with sensor pointing at off-nadir.

# Accuracy of Optical Depth (US Standard Atm., IR Ice Cloud, Edge)



# Precision of Optical Depth (US Standard Atm., IR Ice Cloud, Edge)

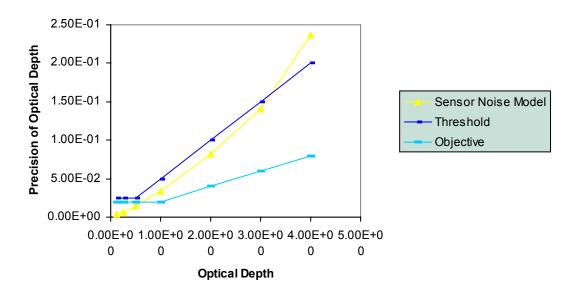


Figure 38. Accuracy and Precision of retrieved optical depths from the IR algorithm for cirrus cloud in US Standard Atmosphere with sensor pointing at edge-of-scan.

#### Cirrus cloud in US Standard Atmosphere, sensor pointing at edge-of-scan

Figure 38 shows the accuracy and precision of retrieved optical depths from the nighttime IR algorithm as a function of optical depths. These results are based on retrievals of combinations of

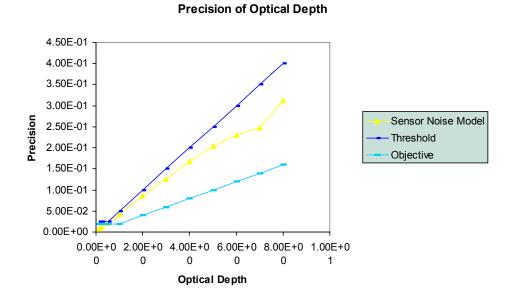
optical depths between 0.125 and 5 and of mean effective sizes between 20 and 100 µm. The straight lines denote threshold (greater of 10 percent or 0.05) and objective (greater of 5 percent or 0.025) values. The accuracy of the retrieved optical depths meets the threshold requirement for optical depths up to 3. The accuracy does not meet the threshold for optical depths larger than 3, mainly due to the exponential relationship between the IR emissivity and the optical depth. For optically thick clouds, the IR emissivity approaches one, so that a small error in the retrieved emissivity can lead to a large error in optical depth. This behavior is very different from the results of the retrieval of the cirrus cloud optical depth using the solar algorithm. The precision of retrieved optical depths meets the threshold (greater of 5 percent or 0.025) requirement for optical depths up to 5. Precision becomes less satisfactory for larger optical depths, because sensitivity of emissivity toward optical depth is reduced due to the exponential relationship between the emissivity and optical depth.

#### Cirrus cloud in tropical atmosphere, sensor pointing at nadir

Figure 39 shows the accuracy and precision of retrieved optical depths from the nighttime IR algorithm as a function of optical depth. These results are based on retrievals of combinations of optical depths between 0.125 and 8 and of mean effective sizes between 20 and 100 µm. Models 1 - 5 meet the accuracy threshold (greater of 10 percent or 0.05) requirement for optical depths up to 8. The straight lines denote threshold (greater of 10 percent or 0.05) and objective (greater of 5 percent or 0.025) values. The accuracy of the retrieved optical depths meets the threshold requirement for optical depths up to 8. The accuracy does not meet the threshold for optical depths larger than 8, mainly because the exponential relationship between the IR emissivity and the optical depth. For optically thick clouds, the IR emissivity approaches one, so that a small error in the retrieved emissivity can lead to a large error in derived optical depth. This behavior is very different from the results of the retrieval of the cirrus cloud optical depth using the solar algorithm. These results are based on retrievals of combinations of optical depths between 0.125 and 7 for the mean effective size of 41.5 um. The precision of retrieved optical depths meets the threshold (greater of 5 percent or 0.025) requirement for optical depths up to 6. Precision becomes less satisfactory for larger optical depths, because sensitivity of emissivity toward optical depth is reduced due to the exponential relationship between the emissivity and optical depth.

#### Water cloud in US Standard Atmosphere, sensor pointing at nadir

Figure 40 shows the accuracy and precision of retrieved optical depths from the nighttime IR algorithm as a function of optical depth for the no-noise case. These results are based on retrievals of combinations of optical depths between 0.125 and 10 and of effective particle sizes between 2 and 12 µm. Two straight lines denote threshold (greater of 10 percent or 0.05) and objective (greater of 5 percent or 0.025) values. The accuracy of the retrieved optical depths meets the threshold requirement for optical depths up to 5. For optical depths larger than 6, the IR emissivity becomes less sensitivity to the optical depth. For optically thick clouds, the IR emissivity approaches one, so that a small error in the retrieved emissivity can lead to a large error in optical depth. This behavior is very different from the results of the retrieval of the water cloud optical depth using the solar algorithm The precision of retrieved optical depths meets the threshold (greater of 5 percent or 0.025) requirement for optical depths up to 5. Precision becomes less satisfactory for larger optical depths, because of lack of sensitivity of emissivity toward optical depth.



#### **Accuracy of Optical Depth**

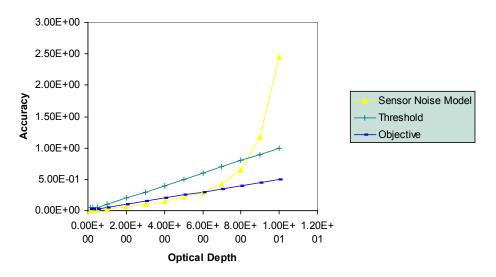
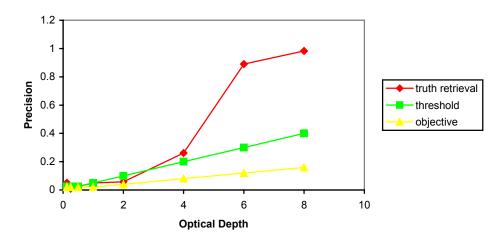


Figure 39. Accuracy and Precision of retrieved optical depths from the IR algorithm for cirrus cloud in Tropical Atmosphere with sensor pointing at edge-of-scan.

#### **Precison of Nighttime IR Water COD**



#### Accuracy of nighttime IR water COD

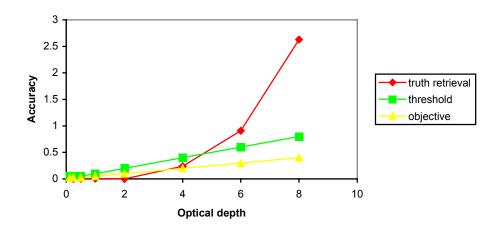


Figure 40. Accuracy and Precision of retrieved optical depths from the IR algorithm for water cloud in US Standard Atmosphere with sensor pointing at nadir for no-noise retrievals.

#### 3.4.1.3 Recommendations

- For daytime retrievals, the sensor noise model is generally suitable for retrieving cloud optical depth using the 0.645, 1.6, and 2.13 µm channels.
- For nighttime retrievals using the IR algorithm, the sensor noise model generally suitable for retrieving cloud optical depth using the 3.75 and 10.8 µm channels.

# 3.4.2 Error Budget Studies

### 3.4.2.1 Algorithm Specific Errors in the Cloud Module

Algorithm specific errors, otherwise referred to as "intrinsic errors", for the cloud modules are discussed below. We define intrinsic errors as errors internal to the algorithms, related to algorithm assumptions or simplifications. Table 6.2-1 of the Error Budget Document gives various intrinsic error sources and their impacts on the retrievals. At this time, only truth retrieval errors are considered. Most of the other intrinsic errors have not been included in the EBs.

### 3.4.2.2 Input Error Sources

There are a number of input parameters to the cloud algorithms. We have attempted to quantify the impact of errors in these input data sources on retrieval accuracy. Our approach has been to develop partial derivatives that quantify the change in a metric value (e.g., measurement accuracy metric, precision metric, or uncertainty metric) to a small perturbation in the input parameter. These partial derivatives are then multiplied by a "standard error", which captures the anticipated rms error in that input parameter. The product of the partial derivative with the standard error gives an estimate of an input parameter's contribution to the error budget for a given EDR. These error budgets have been developed by applying the retrieval algorithms to scenes. In general, these scenes cover approximately a 100 km x 100 km area. The spatial distribution of clouds within the area is determined through the use of TASC's Cloud Scene Simulation Model (CSSM). Processing of CSSM data yields a 2-D distribution of visible cloud optical thickness at 0.1km spatial resolution for We assert an atmospheric scenario that includes sensor geometry, surface a cloud layer. characteristics, atmospheric profiles and cloud base height, cloud top height, and effective particle size. These atmospheric characteristics are used to develop RT look-up tables (LUTs) for the scenario of VIIRS channel radiances as a function of cloud effective particle size and optical depth. These RT LUTs are used to create radiance images for the scene for each VIIRS channel used by the cloud algorithms, by associating a radiance value from the LUT with the optical depth value associated with each 0.1 km data point. The 0.1 km data are aggregated to the VIIRS pixel size. Perturbations to radiances, such as BBR, MTF, geolocation, and calibration are applied to the 0.1 km prior to aggregation. Perturbations to scenario parameters are effected by re-computing the radiance LUTs. The perturbed radiances are then applied to the scene and the unperturbed radiances are used to develop so-called truth retrievals. The retrievals using the perturbed radiances are used to compute SRD metrics and the partial derivatives. This methodology was used to develop EBs for all cloud EDRS, except CBH and CCL. The techniques used for the CBH and CCL EDRs are described in their respective sections.

Table 6.3-1 of the Error Budget Document provides a general list of input parameters used by the cloud algorithms. A given retrieval algorithm may not use all input parameters. The input parameters are divided into three groups: surface parameters, atmospheric profiles, and EDRs and SDRs. Note that error contributions for the VIIRS Cloud Mask (VCM) and Cloud Phase algorithms will be introduced once these models have been tuned for the new bands. Previous experience with these models suggests that the impact on the error budgets should be small.

#### 3.4.2.3 Sensor Error Sources

Several attributes of the sensor were assumed throughout the cloud EDR EBs. These are listed below:

- Sensor Noise Model
- MTF Model: 5
- Band-to-Band Registration (BBR): 0.2 pixel shift
- Geolocation: 0.2 pixel shift
- Absolute Radiometric Calibration (ARA): 0.4% IR channels, 2% solar channels
- Pixel sizes: nadir (753m, 1160m, and 1600m)

The BBR and geolocation errors and MTF model were applied to the 0.1 km radiance data prior to aggregation. We noted extreme sensitivity to the MTF modeling, especially for optical depth retrievals using the solar channels. We are investigating the possibility of an inconsistency in our simulations with respect to the MTF modeling. In addition, we used Sensor Noise Model #3 due to lack of time. We understand that performance with the new baseline should be better anyway

## 3.4.2.4 Specified and Predicted Performances

The Error Budgets for the COT EDR are divided into four areas consistent with the specification and with the four algorithms used to produce this EDR: daytime water cloud, daytime ice cloud, nighttime ice cloud, and nighttime water cloud. The specified and predicted performances based on Error Budget analyses for these four cloud types are provided in Section 6.7.3 of the Error Budget Document.

#### 3.5 ALGORITHM SENSITIVITY STUDIES ON RETRIEVALS OF EFFECTIVE PARTICLE SIZE

The sensitivity studies on the retrievals of effective particle size were carried out using the cloud optical properties retrieval algorithm under development at UCLA. The algorithm sensitivity studies were carried out using the results generated by the radiative transfer model for seven separate scenarios. These scenarios form the basis for defining the input parameters required by the radiative transfer model. They are:

- (1) Cirrus cloud in US Standard Atmosphere, sensor pointing at nadir.
- (2) Cirrus cloud in US Standard Atmosphere, sensor pointing at off-nadir.
- (3) Cirrus cloud in US Standard Atmosphere, sensor pointing at edge-of-scan.
- (4) Cirrus cloud in Tropical Atmosphere, sensor pointing at nadir.
- (5) Cirrus cloud in Sub-arctic Atmosphere, sensor pointing at nadir.
- (6) Cirrus cloud in Desert Atmosphere, sensor pointing at nadir.
- (7) Water cloud in US Standard Atmosphere, sensor pointing at nadir.

For scenarios 1 through 5, synthetic retrievals for both daytime and nighttime conditions were carried out for combinations of input cirrus optical depths and effective particle sizes. For scenario 7, only synthetic retrievals for daytime conditions were carried out.

The LBLE radiative transfer model described in section 3.3.1.7. was used to generate simulated radiances. The input parameters for each scenario have been given in Table 5. For all scenarios, the solar zenith angle is fixed at 32°, the nominal date and location are based on the VIIRS simulated orbit on 1 May at 40° North. For a given specification of atmospheric profile, sun-sensor geometry, retrieval channel characteristics (including central wavelength, bandwidth and response function), cloud type, altitude and thickness, as well as surface albedo and emissivity, the radiative transfer calculations are performed for a wide range of selected visible optical depths and mean effective ice crystal sizes or water droplet radius. For cirrus clouds, the optical depths chosen are: 0.125, 0.25, 0.5, 1, 2, 3, 4, 5, 6, 7, 8, 9, 10, and 12. The mean effective sizes chosen are: 23.9, 30.4, 41.5, 71, and 93 µm. For water clouds the optical depths selected are: 0.125, 0.25, 0.5, 1, 2, 3, 4, 5, 6, 8, 12, 16, 24, 32, 48, and 64. The mean droplet radii selected are: 2, 3, 4, 6, 8, 12, 16, 24, and 30 µm. Many radiance values are thus produced, and they are further processed to create data sets that are appropriate for the parameter retrieval algorithms and algorithm sensitivity tests.

This section addresses results of the algorithm sensitivity analyses for cloud effective particle sizes using the cirrus and water cloud IR and solar retrieval algorithms, which use VIIRS bands at 3.75 and 10.8 µm for IR algorithms and bands at 0.645, 1.610, and 2.130 µm for solar algorithms, respectively. In the present work, the cirrus cloud retrieval algorithm retrieves mean effective size as defined in Equation 1. The ice crystal mean effective size is in the dimension of width (equivalent to the dimension of diameter for a sphere, a cylindrical column or a circular plate), while according to the VIIRS SRD, the cloud effective particle size is defined in the dimension of radius (or half of the diameter for a sphere, a cylindrical column or a circular plate). We thus define the effective particle

size for ice clouds to be half of the ice crystal mean effective size. Therefore, the uncertainty of cloud effective particle size is also half of that for the ice crystal mean effective size. The water cloud retrieval algorithm retrieves effective droplet radius. We define the effective particle size for water clouds to be equal to the effective droplet radius. Illustrative results of these algorithm sensitivity analyses follow.

#### 3.5.1 SNR Tests

The SNR tests address the impact of SNR on both pixel-level and image-level retrievals. The pixel-level retrievals are performed for a single pixel assuming various combinations of cloud optical depth and effective particle size. The imagery-level retrievals are performed for pixels aggregated (averaged) to VIIRS SRD horizontal cell size (HCS). The required HCS is a function of EDR parameter; threshold and objective values are stated. *In this report, only the pixel-level SNR tests are addressed.* The test results are presented separately for the daytime (solar algorithm) and nighttime (IR) algorithms. To create data sets to support the pixel level SNR test, noise based on the sensor noise model is added directly to the radiance contained in the radiance tables and then retrievals are performed using the noise-added data in the tables. To guarantee sample sizes are sufficient to support tests, noise is randomly added to each radiance value 32 times. In effect, 32 noise perturbed radiance tables are created. Retrievals are performed using each of the 32 noise-added tables and the no-noise table. The results of the retrieval process are then statistically analyzed to compute the metrics described in the SRD (accuracy, precision, etc.).

# 3.5.1.1 Daytime Results

### Cirrus cloud in US Standard Atmosphere, sensor pointing at nadir

Figure 41 shows the accuracy and precision of retrieved mean effective sizes as functions of mean effective size. These results are based on retrievals of combination of all possible optical depths (0 to 10) and mean effective sizes (20 to 100  $\mu$ m). Two lines denote threshold (greater of 10 percent or 4  $\mu$ m) and objective (greater of 5 percent or 2  $\mu$ m) values. The accuracy of retrieved mean effective sizes meets both the threshold and objective requirements for the range of mean effective sizes. The accuracy is about the same magnitude across the range of mean effective size for each noise model. The precision of retrieved mean effective sizes meets the threshold requirement for the full range of mean effective sizes. The small values of accuracy and precision errors indicate that the solar algorithm is reasonably accurate and stable in retrieving mean effective sizes.

To examine the effects of adding the VIIRS 2.13  $\mu$ m channel for the retrieval of cirrus cloud parameters, Figure 42 show the accuracy and precision of retrieved effective particle size as functions of optical depth. These results are based on retrievals of combination of all possible optical depths and mean effective sizes using the display of the 0.645 and 2.13  $\mu$ m reflectance correlation. Two straight lines denote threshold (greater of 10 percent or 4 $\mu$ m) and objective (greater of 5 percent or 2 $\mu$ m) values. The accuracy of retrieved effective particle sizes meets both the threshold and objective requirements for most selected values of effective particle sizes. The precision of retrieved effective particle sizes meets the threshold (greater of 5 percent or 2  $\mu$ m) requirements for mean effective particle sizes between 25 and 70  $\mu$ m. For larger optical depths, both the accuracy and precision errors are close to zero. Figure 42 appears similar to Figure. 41. Thus we expect all retrievals using the 0.645-2.13  $\mu$ m correlations will produce similar results as those using the 0.645-1.6  $\mu$ m correlations.

For the rest of the algorithm studies on the solar retrieval algorithm, we will only show results using the  $0.645-1.6 \mu m$  correlation.

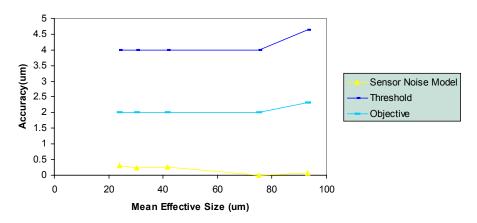
#### Cirrus cloud in US Standard Atmosphere, sensor pointing at off-nadir

Figure 43 shows the accuracy and precision of retrieved mean effective sizes as functions of mean effective size. These results are based on retrievals of combination of all possible optical depths (0 to 10) and mean effective sizes (20 to 100  $\mu$ m). Two lines denote threshold (greater of 10 percent or 4  $\mu$ m) and objective (greater of 5 percent or 2  $\mu$ m) values. The accuracy of retrieved mean effective sizes meets both the threshold and objective requirements for the range of mean effective sizes. The accuracy is about the same magnitude across the range of mean effective size for each noise model. The precision of retrieved mean effective sizes meets the threshold requirement for the full range of mean effective sizes. The precision of retrieved mean effective sizes meets the objective requirement for the full range of mean effective sizes. The small values of accuracy and precision errors indicate that the solar algorithm is reasonably accurate and stable in retrieving mean effective sizes.

#### Cirrus cloud in US Standard atmosphere, sensor pointing at edge-of-scan

Figure 44 shows the accuracy and precision of retrieved mean effective sizes as functions of mean effective size. These results are based on retrievals of combination of all possible optical depths (0 to 10) and mean effective sizes (20 to 100  $\mu$ m). Two lines denote threshold (greater of 10 percent or 4  $\mu$ m) and objective (greater of 5 percent or 2  $\mu$ m) values. The accuracy of retrieved mean effective sizes meets both the threshold and objective requirements for the range of mean effective sizes. The accuracy is about the same magnitude across the range of mean effective size for each noise model. The precision of retrieved mean effective sizes meets the threshold requirement for the full range of mean effective sizes. The precision of retrieved mean effective sizes meets the objective requirement for mean effective sizes between 20 and 80  $\mu$ m. The small values of accuracy and precision errors indicate that the solar algorithm is reasonably accurate and stable in retrieving mean effective sizes.

#### **Accuracy of Mean Effective Size (daytime)**



#### **Precision of Mean Effective Size**

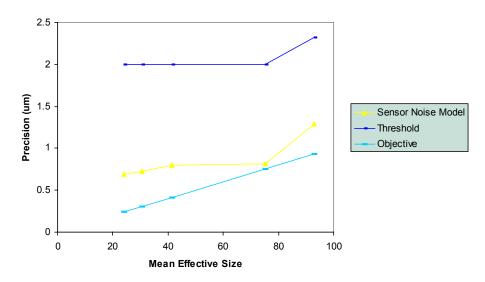
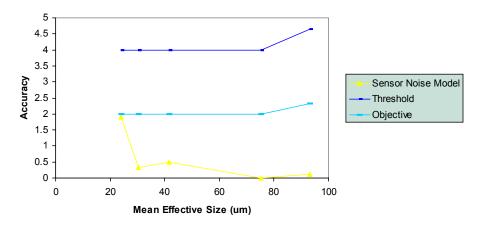


Figure 41. Accuracy and Precision of retrieved effective particle size from the cirrus solar algorithm for US Standard Atmosphere, sensor at nadir, based on the 0.645-1.61 µm correlation.

#### **Accuracy of Mean Effective Size**



#### **Precision of Mean Effective Size**

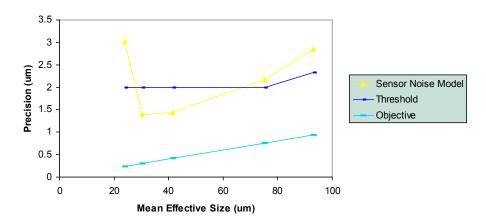
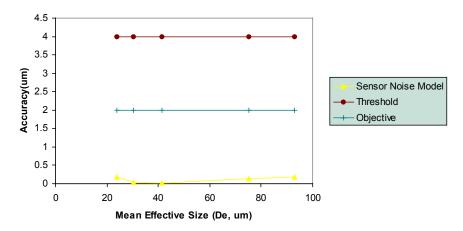


Figure 42. Accuracy and Precision of retrieved effective particle size from the cirrus solar algorithm for US Standard Atmosphere, sensor at nadir, based on the 0.645-2.13  $\mu m$  correlation.

#### **Accuracy of Mean Effective Size**



#### Precision of Mean Effective Size

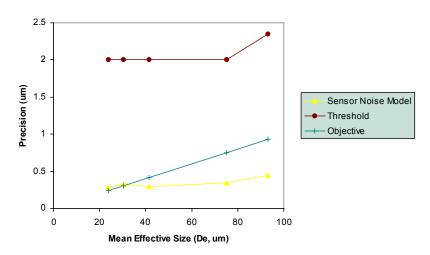
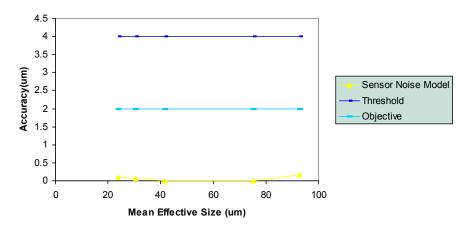


Figure 43. Accuracy and Precision of retrieved effective particle size from the cirrus solar algorithm for US Standard Atmosphere, sensor at off-nadir, based on the 0.645-1.61  $\mu m$  correlation.

#### Accuracy of mean effective size



# Precision of Mean Effective Size, US Standard Atmosphere, ice cloud, solar, edge-of-scan

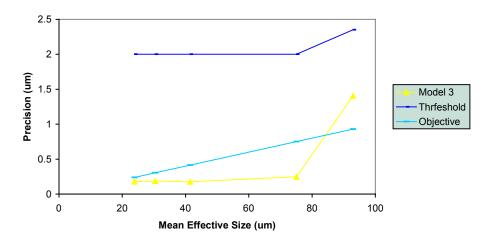


Figure 44. Accuracy and Precision of retrieved effective particle size from the cirrus solar algorithm for US Standard Atmosphere, sensor at edge-of-scan, based on the 0.645-1.61 µm correlation.

# Cirrus cloud in tropical atmosphere, sensor pointing at nadir

Figure 45 shows the accuracy and precision of retrieved mean effective sizes as functions of mean effective size. These results are based on retrievals of combination of all possible optical depths (0 to 10) and mean effective sizes (20 to 100  $\mu$ m). Two lines denote threshold (greater of 10 percent or 4  $\mu$ m) and objective (greater of 5 percent or 2  $\mu$ m) values. The accuracy of retrieved mean effective sizes meets both the threshold and objective requirements for the range of mean effective sizes. The accuracy is about the same magnitude across the range of mean effective size for each noise model. The precision of retrieved mean effective sizes meets the threshold requirement for the full range of mean effective sizes. The precision of retrieved mean effective sizes meets the objective requirement for mean effective sizes between 40 and 100  $\mu$ m. The small values of accuracy and precision errors indicate that the solar algorithm is reasonable accurate and stable in retrieving mean effective sizes.

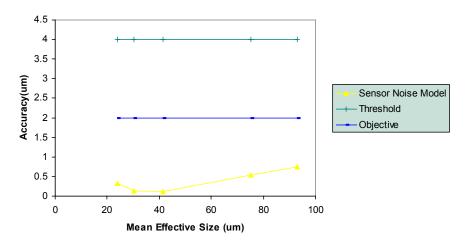
# Cirrus cloud in sub-arctic atmosphere, sensor pointing at nadir

Figure 46 shows the accuracy and precision of retrieved mean effective sizes as functions of mean effective size. These results are based on retrievals of combination of all possible optical depths (0 to 10) and mean effective sizes (20 to 100  $\mu$ m). Two lines denote threshold (greater of 10 percent or 4  $\mu$ m) and objective (greater of 5 percent or 2  $\mu$ m) values. The accuracy of retrieved mean effective sizes meets both the threshold and objective requirements for the range of mean effective sizes. The accuracy is about the same magnitude across the range of mean effective size for each noise model. The precision of retrieved mean effective sizes meets the threshold requirement for the full range of mean effective sizes. The precision of retrieved mean effective sizes meets the objective requirement for the full range of mean effective sizes. The small values of accuracy and precision errors indicate that the solar algorithm is reasonable accurate and stable in retrieving mean effective sizes.

# Cirrus cloud in desert atmosphere, sensor pointing at nadir

Figure 47 shows the accuracy and precision of retrieved mean effective sizes as functions of mean effective size. These results are based on retrievals of combination of all possible optical depths (0 to 10) and mean effective sizes (20 to 100  $\mu$ m). Two lines denote threshold (greater of 10 percent or 4  $\mu$ m) and objective (greater of 5 percent or 2  $\mu$ m) values. The accuracy of retrieved mean effective sizes meets both the threshold and objective requirements for the range of mean effective sizes. The accuracy is about the same magnitude across the range of mean effective size for each noise model. The precision of retrieved mean effective sizes meets the threshold requirement for the full range of mean effective sizes. The precision of retrieved mean effective sizes meets the objective requirement for mean effective sizes between 40 and 100 $\mu$ m. The small values of accuracy and precision errors indicate that the solar algorithm is reasonable accurate and stable in retrieving mean effective sizes.

# Accuracy of r(eff) for Ice Cloud at Nadir for a Tropical Atmosphere



# Precision of r(eff) for Daytime Ice Cloud at Nadir for a Tropical Atmosphere

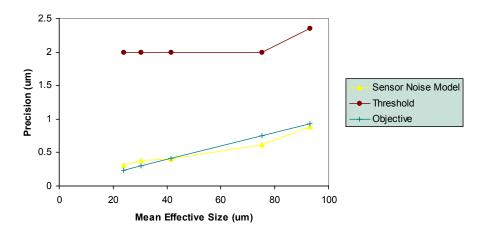
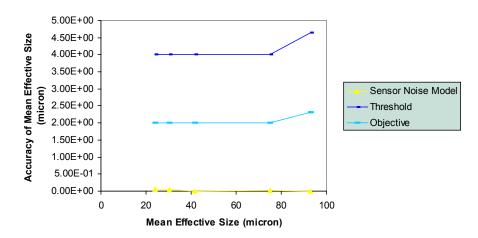


Figure 45. Accuracy and Precision of retrieved effective particle size from the cirrus solar algorithm for Tropical Atmosphere, sensor at nadir, based on the 0.645-1.61 µm correlation.

# Accuracy of Mean Effective Size (Subarctic, Solar Ice Cloud, Nadir)



# Precision of Mean Effective Size (Subarctic, Solar Ice Cloud, Nadir)

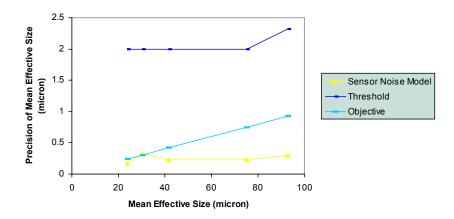
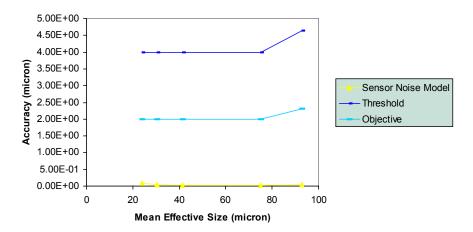


Figure 46. Accuracy and Precision of retrieved effective particle size from the cirrus solar algorithm for Subarctic Atmosphere, sensor at nadir, based on the 1.2-1.61 µm correlation.

# Accuracy of Mean Effective Size (Desert, US Standard Atm., Solar Ice Cloud, Nadir)



#### Precision of Mean Effective Size (Desert, US Standard Atm., Solar Ice Cloud, Nadir)

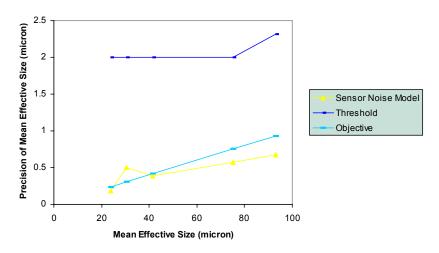


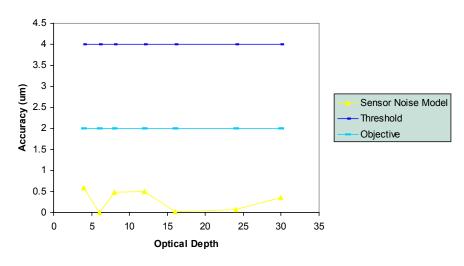
Figure 47. Accuracy and Precision of retrieved effective particle size from the cirrus solar algorithm for Desert Atmosphere, sensor at nadir, based on the 0.645-1.61 µm correlation.

### Water cloud in US Standard Atmosphere, sensor pointing at nadir

Figure 48 shows the accuracy and precision of retrieved droplet effective radius as functions of droplet effective radius. These results are based on retrievals of combination of all possible optical depths (0 to 64) and mean effective sizes (4 to 30  $\mu$ m). For accuracy, the horizontal lines denote threshold (greater of 10 percent or 4  $\mu$ m) and objective (greater of 5 percent or 2  $\mu$ m) values. The accuracy of retrieved droplet effective radius meets both the threshold and objective requirements for the range of droplet effective radius. For precision, the horizontal lines denote threshold (greater of 5 percent or 2  $\mu$ m) and objective (2 percent) values. The precision of retrieved droplet effective radius

meets the threshold requirement for the full range of droplet effective radius. The small values of accuracy and precision errors indicate that the solar algorithm is reasonably accurate and stable in retrieving droplet effective radius.

#### Accuracy of r(eff) for Water Cloud at Nadir



#### Precision of r(eff) for water cloud at nadir

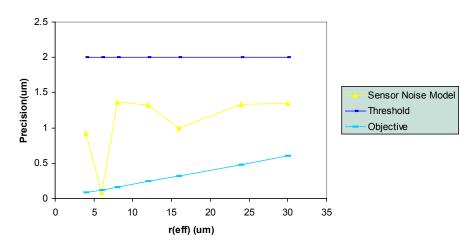


Figure 48. Accuracy and Precision of retrieved effective particle size from the water cloud solar algorithm for US Standard Atmosphere, sensor at nadir, based on the 0.645-1.61 µm correlation.

# 3.5.1.2 Nighttime Results

#### Cirrus cloud in US Standard Atmosphere, sensor pointing at nadir

Figure 49 (a) – (c) shows the accuracy of retrieved mean effective ice crystal sizes from the nighttime IR algorithm as a function of optical depth for three mean effective sizes (23.9, 41.5, and 93 µm. These results are based on retrievals of combinations of optical depths between 0.125 and 7 and of mean effective sizes between 20 and 100 µm. horizontal lines denote threshold (greater of 10 percent or 4 µm) and objective (greater of 5 percent or 2 µm) values. The accuracy of the retrieved mean effective ice crystal size for  $D_e = 23.9 \mu m$  meets the threshold and objective requirements for optical depths up to 7 and 5 respectively. The accuracy of the retrieved mean effective ice crystal size for  $D_e$ = 41.5 µm meets the threshold and objective requirements for optical depths up to 5 and 3 respectively. The accuracy of the retrieved mean effective ice crystal size for  $D_e = 93 \, \mu m$  meets the threshold requirement for optical depths up to 3. The accuracy does not meet the threshold for optical depths larger than 6, mainly due to the indirect effects of the exponential relationship between the IR emissivity and the optical depth. As the emissivity approaches unity, small errors in the retrieved emissivity will be amplified in the retrieved optical depth, and thus the accuracy of the retrieved mean effective size, which depends on the accuracy of the retrieved optical depth, is affected. On the other hand, as the emissivity approaches zero, errors in the retrieved cloud-top temperature increase. Therefore, the accuracy of the retrieved mean effective size, which depends on the accuracy of the retrieved cloud temperature, becomes worse for smaller optical depths. Moreover, the range of optical depths for which the threshold and objective requirements are met becomes smaller for larger ice crystal mean effective size. This is because as the mean effective size increases, the sensitivity of 3.7 µm radiance to mean effective size decreases. Figure 50 shows the precision of retrieved mean effective ice crystal sizes from the nighttime IR algorithm as a function of optical depth for two mean effective sizes (23.9 and 41.5 µm). The horizontal lines denote threshold (greater of 5 percent or 2 um) and objective (2 percent) values. The precision of the retrieved mean effective ice crystal size for  $D_e = 23.9 \,\mu\text{m}$  meets the threshold and objective requirements for optical depths between 1 and 7, and between 5 and 7, respectively. The precision of the retrieved mean effective ice crystal size for  $D_e$  = 41.5 µm meets the threshold for optical depths between 2 and 7.

# Cirrus cloud in US Standard Atmosphere with sensor pointing at off-nadir

Figure 51 (a)–(c) shows the accuracy of retrieved mean effective size from the nighttime IR algorithm as a function of optical depths for three mean effective sizes (23.9, 41.5 and 93 $\mu$ m). These results are based on retrievals of combinations of optical depths between 0.125 and 7 and of mean effective sizes between 20 and 100  $\mu$ m. The horizontal lines denote threshold (greater of 10 percent or 4  $\mu$ m) and objective (greater of 5 percent or 2  $\mu$ m) values. The accuracy of the retrieved mean effective ice crystal size for  $D_e = 23.9 \mu$ m meets the threshold and objective requirements for optical depths up to 10 and between 1 and 5, respectively. The accuracy of the retrieved mean effective ice crystal size for  $D_e = 41.5 \mu$ m meets the threshold and objective requirements for optical depths up to 5 and 3, respectively. The accuracy meets the threshold for  $D_e = 93 \mu$ m and optical depths between 0.5 and 2. As the emissivity approaches unity, small errors in the retrieved emissivity will be amplified in the retrieved optical depth, and thus the accuracy of the retrieved mean effective size, which depends on the accuracy of the retrieved optical depth, is also amplified. On the other hand, as the emissivity approaches zero, errors in the retrieved cloud-top temperature increase. Therefore, the uncertainty of the retrieved mean effective size, which depends on the accuracy of the retrieved cloud temperature,

becomes larger for smaller optical depths. Moreover, the range of optical depths for which the threshold and objective requirements are met becomes smaller for larger ice crystal mean effective size. This is because as the mean effective size increases, the sensitivity of 3.7  $\mu$ m radiance to mean effective size decreases. Figure 52 (a)–(c) shows the precision of retrieved mean effective ice crystal sizes from the nighttime IR algorithm as a function of optical depth for two mean effective sizes (23.9 and 41.5  $\mu$ m). The horizontal lines denote threshold (greater of 5 percent or 2  $\mu$ m) and objective (2 percent) values. The precision of the retrieved mean effective ice crystal size for  $D_e$  = 23.9  $\mu$ m meets the threshold requirement for optical depths between 1 and 10. The precision of the retrieved mean effective ice crystal size for  $D_e$  = 41.5  $\mu$ m meets the threshold for optical depths between 1 and 8. The general pattern of the precision errors as a function of optical depth for various noise models is that it decreases with increasing optical depth. This is possibly because for larger optical depth, both the precision error for cloud temperature and optical depth are small, leading to small precision error values.

#### Cirrus cloud in US Standard Atmosphere with sensor pointing at edge-of-scan

Figure 53 (a)–(c) shows the accuracy of retrieved mean effective size from the nighttime IR algorithm as a function of optical depths for three mean effective sizes (23.9, 41.5 and 93 µm). These results are based on retrievals of combinations of optical depths between 0.125 and 7 and of mean effective sizes between 20 and 100 µm. The horizontal lines denote threshold (greater of 10 percent or 4 µm) and objective (greater of 5 percent or 2 µm) values. The accuracy of the retrieved mean effective ice crystal size for  $D_e = 23.9 \,\mu m$  meets the threshold and objective requirements for optical depths up to 10. The accuracy of the retrieved mean effective ice crystal size for  $D_e = 41.5 \,\mu m$  meets the threshold requirements for optical depths between 1 and 8. The accuracy meets the threshold for  $D_e = 93 \, \mu \text{m}$  for optical depths between 2 and 7. As the emissivity approaches unity, small errors in the retrieved emissivity will be amplified in the retrieved optical depth, and thus the accuracy of the retrieved mean effective size, which depends on the accuracy of the retrieved optical depth, is also amplified. On the other hand, as the emissivity approaches zero, errors in the retrieved cloud-top temperature increase. Therefore, the uncertainty of the retrieved mean effective size, which depends on the accuracy of the retrieved cloud temperature, becomes larger for smaller optical depths. Moreover, the range of optical depths for which the threshold and objective requirements are met becomes smaller for larger ice crystal mean effective size. This is because as the mean effective size increases, the sensitivity of 3.7 um radiance to mean effective size decreases. Figure 54 (a)–(c) shows the precision of retrieved mean effective ice crystal sizes from the nighttime IR algorithm as a function of optical depth for two mean effective sizes (23.9, 41.5 and 93 µm). The horizontal lines denote threshold (greater of 5 percent or 2 μm) and objective (2 percent) values. The precision of the retrieved mean effective ice crystal size for  $D_e = 23.9 \,\mu\text{m}$  meets the threshold requirement for optical depths between 1 and 6. The precision of the retrieved mean effective ice crystal size for  $D_e = 41.5 \,\mu m$  meets the threshold for optical depths between 0.5 and 4. The general pattern of the precision errors as a function of optical depth for various noise models is that it decreases with increasing optical depth. This is possibly because for larger optical depth, both the precision error for cloud temperature and optical depth are small, leading to small precision error values.

# Cirrus cloud in tropical atmosphere, sensor pointing at nadir

Figure 55 (a)–(c) shows the accuracy of retrieved mean effective size from the nighttime IR algorithm as a function of optical depths for three mean effective sizes (23.9, 41.5, and 93 μm). These results

are based on retrievals of combinations of optical depths between 0.125 and 7 and of mean effective sizes between 20 and 100 µm. The horizontal lines denote threshold (greater of 10 percent or 4 µm) and objective (greater of 5 percent or 2 µm) values. The accuracy of the retrieved mean effective ice crystal size for  $D_e = 23.9 \,\mu m$  meets the threshold and objective requirements for optical depths up to 10. The accuracy of the retrieved mean effective ice crystal size for  $D_e = 41.5 \, \mu m$  also meets the threshold and objective requirements for optical depths up to 10. The accuracy of the retrieved mean effective ice crystal size for  $D_e = 93 \mu m$  meets the threshold and objective requirements for optical depths up to 6. The accuracy parameter increases for larger optical depth, mainly due to the indirect effects of the exponential relationship between the IR emissivity and the optical depth. As the emissivity approaches unity, small errors in the retrieved emissivity will be amplified in the retrieved optical depth, and thus the accuracy of the retrieved mean effective size, which depends on the accuracy of the retrieved optical depth, is affected. On the other hand, as the emissivity approaches zero, errors in the retrieved cloud-top temperature increase. Therefore, the accuracy error value of the retrieved mean effective size, which depends on the accuracy of the retrieved cloud temperature, becomes larger for smaller optical depths. Moreover, the range of optical depths for which the threshold and objective requirements are met becomes smaller for larger ice crystal mean effective size. This is because as the mean effective size increases, the sensitivity of 3.7 µm radiance to mean effective size decreases. Overall, for the same optical depth, mean effective size, and noise model, the accuracy for the tropical atmosphere is better than that for the US Standard Atmosphere, because the radiances are more sensitive to the cloud temperature and optical depth.

Figure 56 (a)–(c) shows the precision of retrieved mean effective ice crystal sizes from the nighttime IR algorithm as a function of optical depth for one mean effective size (41.5  $\mu$ m). The horizontal lines denote threshold (greater of 5 percent or 2  $\mu$ m) and objective (2 percent) values. The precision of the retrieved mean effective ice crystal size for  $D_e = 41.5 \mu$ m also meets the threshold for optical depths between 1 and 8. The general pattern of the precision error as a function of optical depth for various noise models is that it decreases down to a certain optical depth value and then increases with increasing optical depth. This is possibly because for intermediate optical depth, the precision errors for both cloud temperature and for optical depth are small, leading to smaller precision error values. But for larger optical depths, the precision error for optical depth is large, leading to larger precision error values.

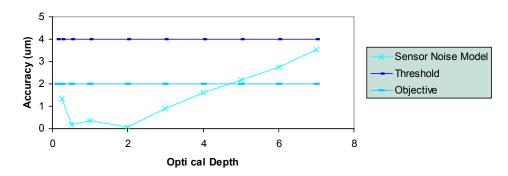
#### Water cloud in US Standard Atmosphere, sensor pointing at nadir

Figure 57 shows the accuracy and precision of retrieved effective particle size from the nighttime IR algorithm as a function of optical depth for the no-noise case. These results are based on retrievals of combinations of optical depths between 0.125 and 10 and of effective particle sizes between 2 and 12  $\mu$ m. The straight lines denote threshold (greater of 10 percent or 4 $\mu$ m) values. The accuracy of the retrieved mean effective sizes meets the threshold requirement for optical depths up to 8. For optical depths larger than 8, the IR emissivity becomes less sensitivity to the optical depth. For optically thick clouds, the IR emissivity approaches one, so that a small error in the retrieved emissivity can lead to a large error in optical depth. This behavior is very different from the results of the retrieval of the water cloud optical depth using the solar algorithm The precision of retrieved mean effective sizes meets the threshold (greater of 5 percent or 2 $\mu$ m) requirement for optical depths up to 7. Precision becomes less satisfactory for larger optical depths, because of lack of sensitivity of emissivity toward optical depth.

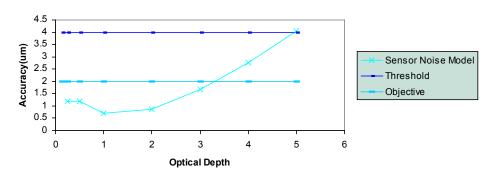
# 3.5.1.3 Recommendations

- For daytime retrievals, the sensor noise model is generally suitable for retrieving cloud effective particle size using the 0.645, 1.6, and 2.13  $\mu m$  channels.
- For nighttime retrievals using the IR algorithm, the sensor noise model is generally suitable for retrieving cloud effective particle size using the 3.75 and 10.8 µm channels.

# Accuracy of Mean Effective Size for De=23.9um



#### Accuracy of Mean Effective Size for De = 41.5 um



# Accuracy of Mean Effective Size for De = 93 um

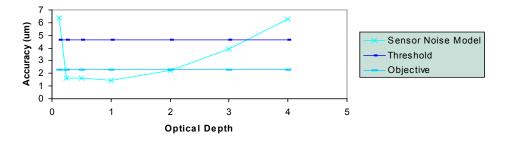
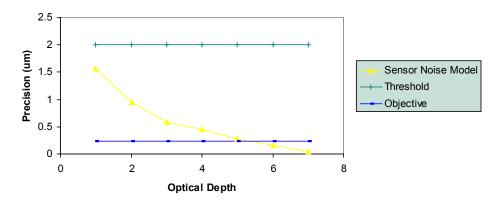


Figure 49. Accuracy of retrieved effective particle size from the cirrus cloud IR algorithm for US Standard Atmosphere, sensor at nadir for  $D_e = 23.9 \, \mu m$ , 41.5  $\mu m$ , and 93  $\mu m$ .

#### Precision of Mean Effective Size for De = 23.9 um



#### Precision of Mean Effective Size for De = 41.5 um

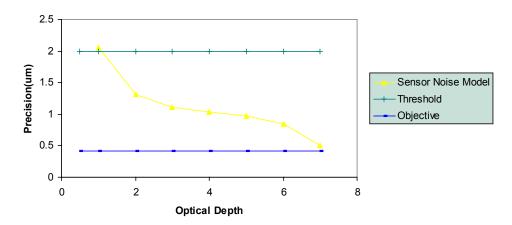
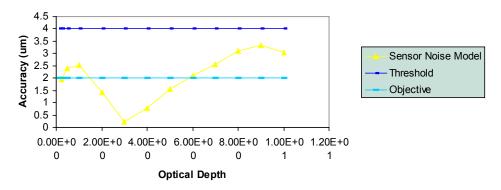
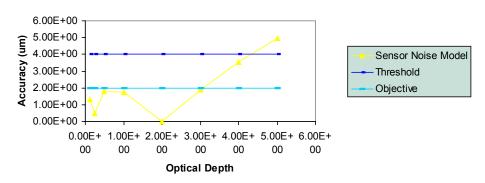


Figure 50. Precision of retrieved effective particle size from the cirrus cloud IR algorithm for US Standard Atmosphere, sensor at nadir for  $D_e = 23.9 \, \mu \text{m}$ , and 41.5  $\mu \text{m}$ .

#### Accuracy of Mean Effective Size for De=23.9um



#### Accuracy of Mean Effective Size for De=41.5 um



# Accuracy of Mean Effective Size for De=93um

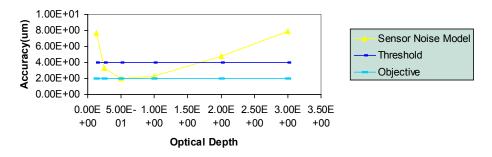
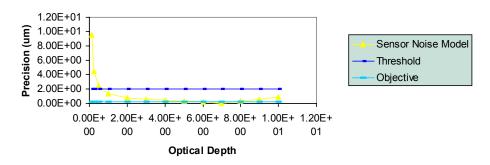


Figure 51. Accuracy of retrieved effective particle size from the cirrus cloud IR algorithm for US Standard Atmosphere, sensor at off-nadir for  $D_e = 23.9 \, \mu \text{m}$ , 41.5  $\mu \text{m}$ , and 93  $\mu \text{m}$ .

#### Precision of Mean Effective Size for De = 23.9um



#### Precision of Mean Effective Size for De = 41.5um

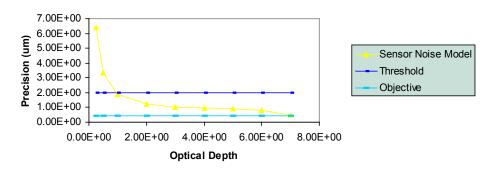
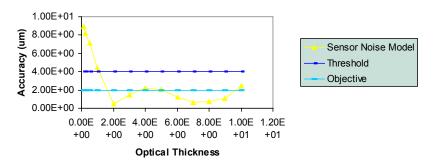
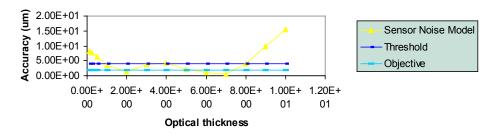


Figure 52. Precision of retrieved effective particle size from the cirrus cloud IR algorithm for US Standard Atmosphere, sensor at off-nadir for  $D_e = 23.9 \, \mu m$ , and 41.5  $\mu m$ .

# Accuracy of Cirrus Mean Effective Size from IR Algorithm for Mid-latitude Edge-of-Scan Scene (De=23.9um)



# Accuracy of Cirrus Cloud Mean Effective Size from IR Algorithm for Mid-latitude Edge-of-Scan Scene (De=41.5um)



# Accuracy of Cirrus Cloud Mean Effective Size from IR Algorithm for Mid-latitude Edge-of-Scan Scene (De=93um)

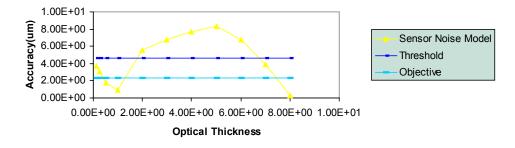
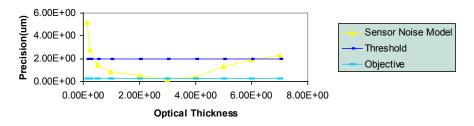


Figure 53. Accuracy of retrieved effective particle size from the cirrus cloud IR algorithm for US Standard Atmosphere, sensor at edge-of-scan for  $D_e = 23.9 \,\mu\text{m}$ , 41.5  $\,\mu\text{m}$ , and 93  $\,\mu\text{m}$ .

# Precision of Cirrus Cloud Mean Effective Size from Nighttime IR Algorithm for a Mid-latitude Edge-of-scan Scene (De = 23.9micron)



# Precision of Cirrus Cloud Mean Effective Size from IR Algorithm for Mid-latitude Edge-of-Scan Scene (De=41.5um)

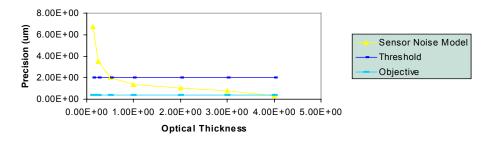
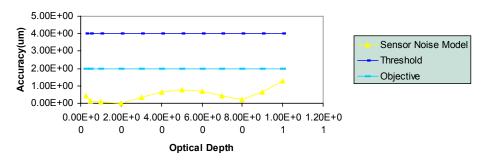
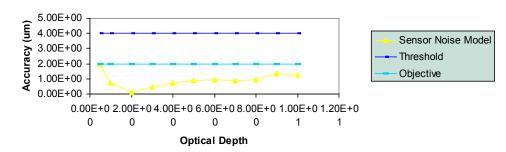


Figure 54. Same as Figure 53, except for Precision.

#### Accuracy of Mean Effective Size (De=41.5um)



#### Accuracy ofMean Effective Size (De=23.9um)



### Accuracy of Mean Effective Size (De=93 um)

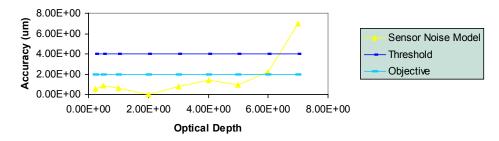


Figure 55. Accuracy of retrieval effective particle size for the cirrus cloud IR algorithm for tropical atmosphere, sensor at nadir for  $D_e = 41.5 \mu m$ , 23.9  $\mu m$ , and 93  $\mu m$ .

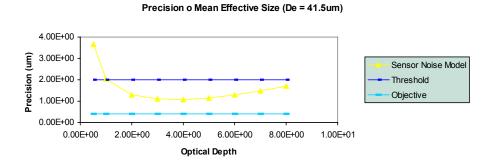
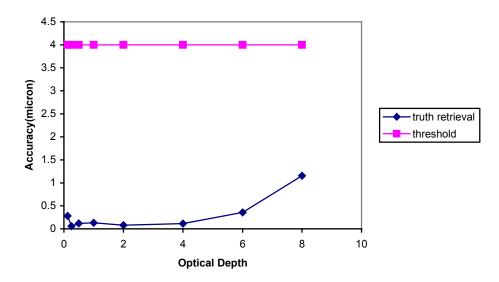


Figure 56. Precision of retrieved effective particle size from the cirrus cloud IR algorithm for Tropical Atmosphere, sensor at nadir for  $D_e = 41.5 \, \mu m$ .

# **Accuracy of Nighttime IR Water EPS**



# **Precision of Nighttime IR Water EPS**

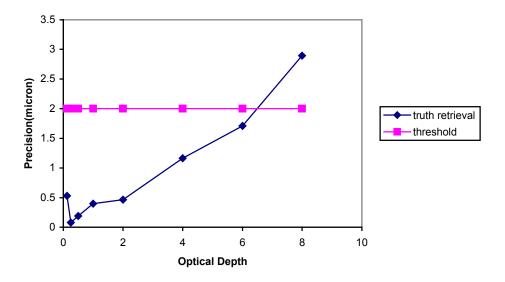


Figure 57. Accuracy and Precision of retrieved effective particle sizes from the IR algorithm for water cloud in US Standard Atmosphere with sensor pointing at nadir for no-noise retrievals.

# 3.5.2 Error Budget Studies

# 3.5.2.1 Algorithm Specific Errors in the Cloud Module

Algorithm specific errors, otherwise referred to as "intrinsic errors", for the cloud modules are discussed below. We define intrinsic errors as errors internal to the algorithms, related to algorithm assumptions or simplifications. Table 6.2-1 of the Error Budget Document gives various intrinsic error sources and their impacts on the retrievals. At this time, only truth retrieval errors are considered. Most of the other intrinsic errors have not been included in the EBs.

# 3.5.2.2 Input Error Sources

112

There are a number of input parameters to the cloud algorithms. We have attempted to quantify the impact of errors in these input data sources on retrieval accuracy. Our approach has been to develop partial derivatives that quantify the change in a metric value (e.g., measurement accuracy metric, precision metric, or uncertainty metric) to a small perturbation in the input parameter. These partial derivatives are then multiplied by a "standard error", which captures the anticipated rms error in that input parameter. The product of the partial derivative with the standard error gives an estimate of an input parameter's contribution to the error budget for a given EDR. These error budgets have been developed by applying the retrieval algorithms to scenes. In general, these scenes cover approximately a 100 km x 100 km area. The spatial distribution of clouds within the area is determined through the use of TASC's Cloud Scene Simulation Model (CSSM). Processing of CSSM data yields a 2-D distribution of visible cloud optical thickness at 0.1km spatial resolution for a cloud layer. We assert an atmospheric scenario that includes sensor geometry, surface characteristics, atmospheric profiles and cloud base height, cloud top height, and effective particle These atmospheric characteristics are used to develop RT look-up tables (LUTs) for the scenario of VIIRS channel radiances as a function of cloud effective particle size and optical depth. These RT LUTs are used to create radiance images for the scene for each VIIRS channel used by the cloud algorithms, by associating a radiance value from the LUT with the optical depth value associated with each 0.1 km data point. The 0.1 km data are aggregated to the VIIRS pixel size. Perturbations to radiances, such as BBR, MTF, geolocation, and calibration are applied to the 0.1 km prior to aggregation. Perturbations to scenario parameters are effected by re-computing the radiance LUTs. The perturbed radiances are then applied to the scene. The unperturbed radiances are used to develop so-called truth retrievals. The retrievals using the perturbed radiances are used to compute SRD metrics and the partial derivatives. This methodology was used to develop Ebs for all cloud EDRS, except CBH and CCL. The techniques used for the CBH and CCL EDRs are described in their respective sections.

Table 6.3-1 of the Error Budget Document provides a general list of input parameters used by the cloud algorithms. A given retrieval algorithm may not use all input parameters. The input parameters are divided into three groups: surface parameters, atmospheric profiles, and EDRs and SDRs. Note that error contributions for the VCM and Cloud Phase algorithms will be introduced once these models have been tuned for the new bands. Previous experience with these models suggests that the impact on the error budgets should be small.

#### 3.5.2.3 Sensor Error Sources

Several attributes of the sensor were assumed throughout the cloud EDR EBs. These are listed below:

- Sensor Noise Model: 3
- MTF Model: 5
- Band-to-Band Registration (BBR): 0.2 pixel shift
- Geolocation: 0.2 pixel shift
- Absolute Radiometric Calibration (ARA): 0.4% IR channels (down to a temperature of 270 K; errors become larger for low temperatures), 2% solar channels
- Pixel sizes: nadir (753m, 1160m, and 1600m)

The BBR and geolocation errors and MTF model were applied to the 0.1 km radiance data prior to aggregation. We noted extreme sensitivity to the MTF modeling, especially for optical depth retrievals using the solar channels. We are investigating the possibility of an inconsistency in our simulations with respect to the MTF modeling. In addition, we used Sensor Noise Model #3 due to lack of time. We understand that performance with the new baseline should be better anyway

# 3.5.2.4 Error Budget for Cloud Effective Particle Size

The Error Budgets for the EPS are divided into four areas consistent with the specification and with the four algorithms used to produce this EDR: daytime water cloud, daytime ice cloud, nighttime ice cloud, and nighttime water cloud. The specifications and predicted performances based on Error Budget Analyses for these four cloud types are provided in Section 6.6.3 of the Error Budget Document.

#### 3.6 PRACTICAL CONSIDERATIONS

### 3.6.1 Numerical Computation Considerations

The retrieval algorithms shall be convertible into operational code that is compatible with a 20 minute maximum processing time. This maximum processing time is allocated not only to the cloud retrieval, but also to the retrieval of other VIIRS EDRs. The estimated processing time of this retrieval system is given as follows. For off-line processing, the generation of radiance tables would take a few CPU hours for all the cloud-retrieval channels and for each scenario. For on-line processing, the solar algorithm (< 1 millisecond per retrieval) will be faster than the IR algorithm (~ 10 times the solar algorithm). The solar algorithm would easily meet the operational timeline for pixel-by-pixel retrievals. Whether the IR algorithm can meet the operational timeline remains to be investigated.

In regard to the data storage, the following storage estimates are based on the simulated VIIRS orbit information. The sensor scanning rate is assumed to be 0.677 sec/scan. In one circular orbit (North Pole to North Pole), there are 8900 scan lines. Thus the period of the sensor is  $0.677 \times 8900 = 6025$  sec or 100.4 minutes. With 3000 pixels in each scan line, the total pixel number in one day's flight is  $3000 \times 8900 \times 1440 \times 1400 \times 1000 \times 1000$ 

The cloud effective particle size has values ranging between 0 to 100  $\mu$ m for cirrus clouds, and between 0 to 50  $\mu$ m for water clouds. The parameter value can be scaled by 10 to achieve necessary accuracy. Thus, it takes 10 bits to store the integer part of the parameter. Thus the data storage required per day is approximately 3.8 x 10  $^8$  pixels x 10 bits / pixel = 3.8 x 10  $^9$  bits = 475 Mbytes.

The cloud optical depth has values ranging from 0 to 10 for cirrus clouds, and from 0 to 64 for water clouds. The parameter value can be scaled by 10 to achieve necessary accuracy. Thus, it takes 6 bits to store the integer part of the parameter. Thus the data storage required per day is approximately 3.8 x  $10^8$  pixels x 6 bits / pixel =  $2.28 \times 10^9$  bits = 285 Mbytes.

# 3.6.2 Programming and Procedural Considerations

For solar retrieval algorithms, comprehensive look-up tables of reflectances as functions of sunsensor geometry, surface albedo, atmospheric sounding, cloud EPS, and COT will be developed for retrieval of cloud properties in almost all atmospheric and surface conditions. For IR retrieval algorithms, parameterization of extinction ratio in terms of EPS and COT for all possible atmospheric conditions will be executed.

# 3.6.3 Configuration of Retrievals

Configuration of the retrieval algorithms will be developed to eliminate the need to hard-wire specific parameters (e.g. clear radiances, atmospheric and surface type and sensor zenith angle).

# 3.6.4 Quality Assessment and Diagnostics

The assessment of the quality of retrievals will fall into four categories: Sensor Parameters; Environmental Scenario; Cloud Scenario; and Ancillary Data. Experience gained through simulations, and eventually by validation, will be captured and used to assess the quality of retrievals and provide guidance to the users of these products in the form of data quality flags. A list of parameters or situations that may influence data quality follows.

- Sensor Parameters. The qualities of sensor data include:
  - Sensor noise.
  - Radiance calibration.
  - Geolocation.
  - MTF
  - Band-to-Band registration.
- Environmental Scenario. Particulars of the environmental scenario that may affect retrieval accuracy include:
  - Values of Environmental Parameters. Sensitivity studies and flow down indicate that retrieval accuracy is a function of the particular values of environmental parameters.
  - Atmospheric inversion identified in sounding.
- *Cloud Scenario*. The qualities or values of other cloud parameters that may affect retrieval accuracy include:

- Cloud optical depth. Flow down results show that retrieval accuracy can be a function of optical depth.
- Cloud effective particle size. Flow down results show that retrieval accuracy can be a function of effective particle size.
- Existence of multilayer clouds. Multilayer clouds are difficult to identify and have an impact on radiance measurements. Therefore, they will affect retrievals when a single layer cloud is assumed in the radiative transfer analysis or retrieval algorithm. Flow down results show some sensitivity to satellite view geometry.
- Solar position. Solar position influences UCLA IR cirrus parameter retrievals during daytime.
- Ancillary Data
  - In general, the quality of ancillary data affects the quality of retrievals.

Based on results from error budget studies, quality flags will be assigned to cloud property products.

# 3.6.5 Exception Handling

Clear and cloud-edge pixels identified by the cloud mask will not be subject to the retrieval algorithm. Pixels with corrupted data will be scanned and filtered out.

#### 3.7 ALGORITHM VALIDATION

Although the above algorithm sensitivity studies have demonstrated the applicability of remote sensing methods to the determination of cloud optical and microphysical properties, more experimental studies are required in order to assess the robustness and accuracy of these methods when applied to measurements on a global scale. Validation efforts are required in order to assess the validity of these methods, since many factors affect the successful retrieval of these parameters when applied to real data in a realistic atmosphere (e.g., Ou *et al.*, 1995). The following is a list of available data for validation

#### a. MODIS data

MODIS (Moderate Resolution Imagining Spectroradiometer) is a scanning spectro-radiometer on board the polar orbiting TERRA(EOS/AM-1) satellite which was launched in December 1999. It possesses 36 channels ranging from 0.412μm to 14.24μm. In accordance with cirrus detection and retrieval the 0.645μm band (channel 1), 0.858μm band (channel 2), 1.240μm band (channel 5), 1.375μm band (channel 26), 1.640μm band (channel 6), 2.13μm band (channel 7), 11.03μm band (channel 31), 12.02μm band (channel 32), and the 13.64μm band (channel 34) will be used. Channels 1, 2, 5, 6, and 7 have 500m resolution while the others have 1 km resolution. Level 1-B data is presently available.

#### b. AVHRR data

The AVHRR (Advanced Very-High Resolution Radiometer) scanning radiometer exists on board the NOAA 10, 14, and 15 satellites collecting data from five channels:  $0.63\mu m$ ,  $0.90\mu m$ ,  $3.7\mu m$ ,  $10.9\mu m$ , and  $12.0\mu m$ .

#### c. ARM data

A vertically pointing 35 GHz Doppler radar along with a multi-pulsed lidar is operational at the Southern Great Plains Atmospheric Radiation Measurement (SGP ARM) site located in northern Oklahoma. Backscattered radiation from the radar can resolve cloud information to 90m in the vertical direction up to a range of 20km. Three Intensive Observation Periods (IOP) were made in conjunction with the overfly of an AVHRR instrument and in-situ measurements from aircraft: April 2-22, 1997; September 15-October 4, 1997; and April 27-May 17, 1998. In addition, surface polarization diversity lidar measurements in the 1998 IOP were made. Future IOP's are expected to take place in coordination with MODIS overflies.

#### d. Microphysical data

In order to obtain a realistic representation of the particle size and shape observed in natural ice clouds, the mean effective size must be obtained. This parameter can be thought of as the number density weighted volume of the ice crystals divided by their number density weighted perpendicular cross sections averaged over the entire vertical column. During the Cloud IOP's at the SGP ARM site, the University of North Dakota Citation and King Air performed in-situ microphysical measurements. These data are essential in the computation of the look-up tables and for the validation of values retrieved from satellite observations.

#### 3.8 ALGORITHM DEVELOPMENT SCHEDULE

During the past two and a half years, the cloud property retrieval algorithms have been mainly used to perform sensitivity studies, including SNR tests and Error Budget studies. It is expected that the development of these algorithms will be completed within the next two to three years.

# 4.0 ASSUMPTIONS AND LIMITATIONS

# 4.1 ASSUMPTIONS

There are several assumptions involved in the theoretical development of retrievals of cloud effective particle size and optical depth:

- For the retrieval of ice cloud parameters, representative ice crystal size distributions are based on *in situ* observations obtained during field experiments over mid-latitude areas. Ice crystals are assumed to be randomly oriented, and the ice crystal habits considered are solid columns and plates.
- For the retrieval of water cloud parameters, a gamma function is assumed to be the typical droplet size distribution, with water droplets assumed to be spherical in shape.

#### 4.2 LIMITATIONS

There are also a few limitations of the current algorithms:

- The solar and IR algorithms are applicable for a single-layer, cirrus or water cloud. An IR retrieval algorithm for inferring multi-layer cirrus/water cloud parameters has been developed (Ou et al., 1998). The accuracy and precision of this algorithm remains to be investigated. The solar retrieval algorithm for inferring multi-layer cirrus/water cloud parameters remains to be developed.
- A comprehensive look-up table for the solar algorithm is yet to be developed. This look-up table will contain reflectance values for all possible ranges of sun-sensor geometry, optical depth, mean effective size, and surface albedos.
- Radiative transfer within the cloud is assumed to be plane-parallel.

# 5.0 REFERENCES

- Ambartzumian, V.(1936). The effects if the absorption lines on the radiative equilibrium of the outer layers of the stars. *Publ. Obs. Astron.* Univ. Leningrad, 6, 7-18.
- Anderson, G. P., R. H. Picard, and J. H. Chetwynd (1995). Proceedings of the 17<sup>th</sup> Annual Review Conference on Atmospheric Transmission Models. Special Report No. 274, Phillips Laboratory/Geophysics Directorate, MA. May 1995, 332 pp.
- Arking, A. and J. D. Childs (1985). Retrieval of cloud cover parameter from multi-spectral satellite images. *J. Clim. Appl. Meteor.*, 23, 322-333.
- Arking, A., and K. Grossman (1972). The influence of line shape and band structure on temperatures in planetary atmospheres. *J. Atmos. Sci.*, 29, 937-949.
- Auer, A. H., Jr. and D. L. Veal (1970). The dimension of ice crystals in natural clouds. *J. Atmos. Sci.*, 27, 919-926.
- Baum, B. A., D. P. Kratz, P. Yang, S. C. Ou, Yongxiang Hu, P. Soulen, S. C. Tsay (2000) Remote sensing of cloud properties using MODIS Airborne simulator imagery during SUCCESS. I. Data and models. *J. Geophys. Res.* (in press)
- Berk, A., L. S. Bernstein, and D. C. Robertson (1989). MODTRAN: A Moderate Resolution Model for LOWTRAN 7. Report AFGL-TR-89-0122 (Air Force Geophysics Laboratory), Hanscom AFB, MA 01731.
- Chylek, P., P. Damiano, N. Kalyaniwalla, and E. P. Shettle (1995). Radiative properties of water clouds: Simple approximations. *Atmos. Res.*, 35, 139-156.
- Clothiaux, E. E., et al. (1995). An evaluation of a 94-GHz radar for remote sensing of cloud properties. J. Atmos. Oceanic Technol., 12, 201-229.
- Curran, R. J., H. L. Kyle, L. R. Blaine, J. Smith, and T. D. Clem (1981). Multichannel scanning radiometer for remote sensing cloud physical parameters. *Rev. Sci. Instrum.*, 52, 1546-1555.
- d'Almeida, G. A., P. Koepke, and E. P. Shettle (1991). Atmospheric Aerosols, Global Climatology and Radiative Characteristics. A. Deepak Publishing, Hampton, VA., 561pp.
- Fu, Q. and K. N. Liou (1992). On the correlated k-distribution method for radiative transfer in nonhomogeneous atmospheres. *J. Atmos. Sci.*, 49, 2139-2156.
- Fu, Q. and K. N. Liou (1993). Parameterization of the radiative properties of cirrus clouds. *J. Atmos. Sci.*, 50, 2008-2025.
- Hansen, J. E. and J. B. Pollack (1970). Near-infrared light scattering by terrestrial clouds. *J. Atmos. Sci.*, 27, 265-281.
- Hansen, J. E., and L. D. Travis (1974). Light scattering in planetary atmospheres. *Space Sci. Rev.*, 16, 527-610.

- Heymsfield, A. J. and C. M. R. Platt (1984). A parameterization of the particle size spectrum of ice clouds in terms of the ambient temperature and ice water content. *J. Atmos. Sci.*, 41, 846-855.
- Isaacs, R. G., W.-C. Wang, R. D. Worsham, and S. Goldenberg (1986). Multiple scattering Treatment for use in the LOWTRAN and FASCODE models. Report AFGL-TR-88-0073 (Air Force Geophysics Laboratory), Hanscom AFB, MA 01731.
- Isaacs, R. G., W.-C. Wang, R. D. Worsham, and S. Goldenberg (1987). Multiple scattering LOWTRAN and FASCODE models. *Appl. Opt.*, 26, 1272-1281.
- Kneizys, F. X., G. P. Anderson, E. P. Shettle, and Coauthors (1988). User's guide to LOWTRAN 7. Air Force Geophysics Lab., Hanscom Air Force Base, MA., 136 pp.
- King, M. D., W. P. Menzel, P. S. Grant, J. S. Myers, G. T. Arnold, S. E. Platnick, L. E. Gumley, S. C. Tsay, C. C. Moeller, M. Fitzgerald, K. S. Brown, and F. G. Osterwisch, (1996). Airborne scanning spectrometer for remote sensing of cloud, aerosol, water vapor and surface properties, *J. Atmos. Oceanic Technology*, 13, 777-794.
- King, M. D., S.-C. Tsay, S. E. Platnick, M. Wang, and K. N. Liou, (1996). Cloud retrieval algorithms for modis: Optical thickness, effective particle radius, and thermodynamic phase. MODIS Algorithm Theoretical Basis Document, No. ATBD-MOD-05.
- Kratz, D. P., (1995) The correlated k-distribution technique as applied to the AVHRR channels. *J. Quant. Spectrosc. Radiat.* Transfer, 53, 501-517.
- Larsen, N. (1994). Improved multiple scattering computations in a moderate resolution radiative transfer model. Geophysical Institute Report, UAG-R-321, U of Alaska, Fairbank, 168pp.
- Liou, K. N. (1992). Radiation and cloud processes in the atmosphere: Theory, observation and modeling. Oxford University Press, 487 pp.
- Liou, K. N., and Y. Takano (1994). Light scattering by nonspherical particles: Remote sensing and climatic implications. *Atmos. Res.*, 31, 271-298.
- Liou, K. N., S. C. Ou, Y. Takano, F. P. J. Valero, and T. P. Ackerman (1990). Remote sounding of the tropical cirrus cloud temperature and optical depth using 6.5 and 10.5 μm radiometers during STEP. *J. Appl. Meteor.*, 29, 715-726.
- Liou, K. N., P. Yang, Y. Takano, K. Sassen, T. Charlock, and W. Arnott (1998). On the radiative properties of contrail cirrus. *Geophys. Res. Lett.*, 25, 1161-1164.
- Minnis, P., D. P. Garber, D. F. Young, R. F. Arduini, and Y. Takano (1998). Paramerizations of reflectance and effective emittance for satellite remote sensing of cloud properties. *J. Atmos. Sci.*, 55, 3313-3339.
- Nakajima, T. and M. D. King (1990). Determination of the optical thickness and effective particle radius of clouds from reflected solar radiation measurements. Part I: Theory. *J. Atmos. Sci.*, 47, 1878-1893.

- Ou, S. C., K. N. Liou, W. M. Gooch, and Y. Takano (1993). Remote sensing of cirrus cloud parameters using advanced very-high-resolution radiometer 3.7- and 10.9 μm channels. *Appl. Opt.*, 32, 2171-2180.
- Ou, S. C., K. N. Liou, Y. Takano, N. X. Rao, Q. Fu, A. J. Heymsfield, L. M. Miloshevich, B. Baum, and S. A. Kinne (1995). Remote sounding of cirrus cloud optical depths and ice crystal sizes from AVHRR data: Verification using FIRE II IFO composite measurements. *J. Atmos. Sci.*, 52, 4143-4158.
- Ou, S. C., K. N. Liou, and B. A. Baum (1996). Detection of multilayer cirrus cloud systems using AVHRR data: Verification based on FIRE II IFO composite measurements. *J. Appl. Meteor.*, 35, 177-191.
- Ou, S. C., K. N. Liou, and T. R. Caudill (1998). Remote sounding of multilayer cirrus cloud systems using AVHRR data collected during FIRE II IFO. *J. Appl. Meteor.*, 37, 241-254.
- Rayleigh, Lord (1918). Light scattering by atmospheric molecules. Phil. Mag., 35, 373-390.
- Rao, N. X., S. C. Ou, and K. N. Liou (1995). Removal of the solar component in AVHRR 3.7-μm radiances for the retrieval of cirrus cloud parameters. *J. Appl. Meteor.*, 34, 481-499.
- Revercomb, H. E., H. Bujis, H. B. Howell, D. D. LaPorte, W. L. Smith, and L. A. Sromovsky (1988). Radiometeric calibration of IR Fourier transform spectrometers: Solution to a problem with the High-spectral resolution Interferometer Sounder. Appl. Opt., 27, 3210-3218.
- Rolland, P., K. N. Liou, M. D. King, S. C. Tsay, and G. M. McFarquhar (2000). Remote sensing of optical and microphysical properties of cirrus clouds using MODIS channels: methodology and sensitivity to physical assumptions. *J. Geophys. Res.*, (in press).
- Spinhirne, J. D., R. Boers and W. D. Hart (1989). Cloud top liquid water from lidar observations of marine stratocumulus. *J. Appl. Meteor.*, 28, 81-90.
- Stamnes, K., S.-C. Tsay, W. Wiscombe, and K. Jayaweera (1988). Numerically stable algorithm for discrete-ordinate method radiative transfer in multiple scattering and emitting layered media. *Appl. Opt.*, 27, 2502-2509.
- Szejwach, G. (1982). Determination of semi-transparent cirrus cloud temperature from infrared radiances: Application to METEOSAT. *J. Appl. Meteor.*, 21, 384-393.
- Takano, Y., and K. N. Liou (1989). Solar radiative transfer in cirrus clouds. Parts I and II. *J. Atmos. Sci.*, 46, 3-19, 20-36.
- Takano, Y., and K. N. Liou (1993). Transfer of polarized infrared radiation in optically anisotropic media: application to horizontally oriented ice crystals. *J. Opt. Soc. Am. A*, 10, 1243-1256.
- Takano, Y., and K. N. Liou (1995). Radiative transfer in cirrus clouds. Part III: Light scattering by irregular ice crystals. *J. Atmos. Sci.*, 52, 818-837.

- Takano, Y., K. N. Liou, and P. Minnis (1992). Effects of small ice crystals on cirrus infrared radiative properties. J. Atmos. Sci.,49, 1487-1493.
- Thekaekara, M. P. (1976). Solar irradiance: total and spectral and its possible variations. *Appl. Opt.*, 15, 915-920.
- Twomey, S., and T. Cocks (1982). Spectral reflectance of clouds in the near-infrared: Comparison of measurements and calculations. J. Meteor. Soc. Japan, 60, 583-592.
- Twomey, S., and T. Cocks (1989). Remote sensing of cloud parameters from spectral reflectance in the near-infrared. Beitr. Phys. Atmos., 62, 172-179.
- Yang, P., and K. N. Liou (1995). Light scattering by hexagonal ice crystals: comparison of finite-difference time domian and geometric optics models. *J. Opt. Soc. Am. A.*, 12, 162-176.
- Yang, P., and K. N. Liou (1996a). Geometric-optics-integral-equation method for light scattering by nonspherical ice crystals. *Appl. Opt.*, 35, 6568-6584.
- Yang, P., and K. N. Liou (1996b). Finite-difference time domain method for light scattering by small ice crystals in three-dimensional space. *J. Opt. Soc. Am. A*, 13, 2072-2085.

# APPENDIX A. SRD REQUIREMENTS

This Appendix describes the threshold and objective requirements for cloud effective particle size and optical depth. The following tables are adopted from the VIIRS SRD.

# Cloud Optical Thickness (IORD Name: Cloud Optical Depth/Transmissivity) (Paragraph 40.4.6)

Cloud optical depth is defined as the extinction (scattering + absorption) vertical optical thickness of all cloud layers in a vertical column of the atmosphere. Optical thickness ( $\tau$ ) is related to transmittance (t) by t= exp (- $\tau$ ). Optical thickness is wavelength dependent and is to be measured in at least two narrow bands centered at 450 nm and 850 nm, with TBD nm bandwidth. Table A2 shows the thresholds and objectives of the parameters relevant to the retrieval of cloud optical depth.

**Units**: Dimensionless

Table A1. Thresholds and objectives of the cloud optical depth EDR parameters.

Para. No.		Thresholds	Objectives
	a. Horizontal Cell Size		
40.4.6-10	Moderate, worst case	50 km (TBR)	10 km
40.4.6-11	2. Fine, at nadir	1 km (TBR)	(TBD)
40.4.6-2	b. Horizontal Reporting Interval	(TBD)	(TBD)
40.4.6-3	c. Horizontal Coverage	Global	Global
40.4.6-4	d. Measurement Range	0 - 10 optical depth	(TBD)
40.4.6-5	e. Measurement Accuracy (moderate HCS product)	greater of 10 % or 0.05 optical depth	greater of 5 % and (TBD)
40.4.6-6	f. Measurement Precision (moderate HCS product)	greater of 5 % or 0.025 optical depth	greater of 2 % and (TBD)
40.4.6-12	e. Measurement Uncertainty (fine HCS product)	greater of 10 % (TBR) or 0.05 optical depth*	(TBD)
40.4.6-7	g. Long Term Stability	2 percent	1 percent
40.4.6-8	h. Mapping Uncertainty	4 km	1 km
	i. Maximum Local Average Revisit Time	8 hrs	3 hrs
	j. Maximum Local Refresh	(TBD)	(TBD)
40.4.6-9	n. Maximum Swath Width (All other EDR thresholds met)	3000 km (TBR)	(TBD)

<sup>\*</sup> Applies at nadir for the fine product nadir HCS and at any swath location for an HCS that intercepts the same in-track and cross-track angles, measured from the satellite location, as the nadir HCS. The fact that this requirement is expressed in terms of an HCS that grows geometrically across the swath dies not imply that a varying HCS is required or desired in the reported product at either the fine or moderate scales.

SBRS Document #: Y2393 Raytheon

# Cloud Effective Particle Size (Paragraph 40.4.3)

Effective cloud particle size is defined as the ratio of the third moment of the drop size distribution to the second moment, averaged over a layer of air within a cloud. Table A1 shows the thresholds and objectives of the parameters relevant to the retrieval of cloud effective particle size.

Units: µm

Table A2. Thresholds and objectives of the cloud effective particle size EDR parameters.

Para. No.		Thresholds	Objectives
	a. Horizontal Cell Size		
40.4.3-15	Moderate, worst case	50 km (TBR)	10 km
40.4.3-16	2. Fine, at nadir	1 km (TBR)	(TBD)
40.4.3-2	b. Horizontal Reporting Interval	(TBD)	(TBD)
40.4.3-3	c. Vertical Cell Size	Vertical Reporting	Vertical Reporting
		Interval	Interval
40.4.3-4	d. Vertical Reporting Interval	1.0 km(TBR)	0.3 km
40.4.3-5	e. Horizontal Coverage	Global	Global
40.4.3-6	f. Vertical Coverage	0 - 20 km	0 - 30 km
40.4.3-7	g. Measurement Range	0 - 50 μm	(TBD)
40.4.3-8	h. Measurement Accuracy	greater of 10 % or 4 µm	greater of 5 % or 2 µm
	(moderate HCS product)		
40.4.3-9	i. Measurement Precision	greater of 5 % or 2 μm	2 %
	(moderate HCS product)		
40.4.3-17	o. Measurement Uncertainty (fine	greater of 10% (TBR) or	(TBD)
	HCS product)	2 μm (TBR)*	
40.4.3-10	j. Long Term Stability	2 %	1 %
40.4.3-11	k. Mapping Uncertainty	4 km	1 km
	I. Maximum Local Average Revisit	8 hrs	3 hrs
	Time		
	m. Maximum Local Refresh	(TBD)	(TBD)
40.4.3-14	n. Maximum Swath Width (All	3000 km (TBR)	(TBD)
	other EDR thresholds met)		

<sup>\*</sup> Applies at nadir for the fine product nadir HCS and at any swath location for an HCS that intercepts the same in-track and cross-track angles, measured from the satellite location, as the nadir HCS. The fact that this requirement is expressed in terms of an HCS that grows geometrically across the swath dies not imply that a varying HCS is required or desired in the reported product at either the fine or moderate scales.

#### Appendix B. ESTIMATING MEAN CIRRUS CLOUD TEMPERATURE

This section describes an optimization scheme to find a first guess of mean cirrus cloud temperature in the IR algorithm.

Using the VIIRS 3.7 and 10.8  $\mu$ m channel data, we proceed to solve the complete set of retrieval equations: 43 through 45, 49, and 57. The solution method follows that described in Ou *et al.* (1993, 1995, 1998). According to the flow chart shown in Figure 15, we need to make a first guess of  $\tau$ ,  $k_2/k_1$ , and  $D_e$ . For this purpose, we use an optimization method to obtain the best-fit mean cirrus cloud temperature. Subsequently, the first guess of  $\tau$ ,  $k_2/k_1$ , and  $D_e$  may be obtained from Equations 59 through 61. The optimization scheme is divided into two parts. The first part is to find an optimal value of  $k_2/k_1$ , given a cloud temperature  $T_c$ . The second part is to find an optimal value of  $T_c$ . The optimization scheme is described as follows.

In the first part, we rewrite Equation 48 in the form:

$$y = k'x \tag{B.1}$$

where:

$$y = \ln\left[\frac{R_2 - B_2(T_c)}{R_{a2} - B_2(T_c)}\right]$$
 (B.2)

$$x = \ln\left[\frac{R_1 - f[B_2(T_c)]}{R_{c1} - f[B_2(T_c)]}\right]$$
(B.3)

$$k' = k_2 / k_1 \tag{B.4}$$

For a given cloud temperature T<sub>c</sub>, we search for an optimal k' that minimizes the residual:

$$E^{2}(k') = \sum_{i} (y_{i} - k'x_{i})y_{i}$$
(B.5)

where i represents the pixel index. Equation B.5 is a weighted residual expression with weight  $y_i$ . Because for  $R_{2i} \rightarrow B_2(T_c)$ ,  $y_i \rightarrow -\infty$ , and for  $R_{2i} \rightarrow R_{a2}$ ,  $y_i \rightarrow 0$ , more weight is put on those pixels whose radiance is closer to the Planck function at  $T_c$ . Following the typical procedures for optimization for a linear equation, we make:

$$\frac{\partial E^2(k')}{\partial k'} = 0 \tag{B.6}$$

Thus the optimal value of k' can be expressed as:

$$k'_{opt}(T_c) = \frac{\sum_{i} y_i^2 x_i}{\sum_{i} y_i x_i^2}$$
 (B.7)

In the second part, we again rewrite Equation 47 in the following form:

$$R_{1} = f[B_{2}(T_{c})] + \left[\frac{R_{2} - B_{2}(T_{c})}{R_{a2} - B_{2}(T_{c})}\right]^{1/k'_{opt}(T_{c})} \left\{R_{a1} - f[B_{2}(T_{c})]\right\}$$
(B.8)

To obtain an estimate of mean cirrus cloud temperature  $T_c$ , we will minimize the following weighted residual:

$$E_2(T_c) = \sum_{i} \left\{ R_1 - f[B_2(T_c)] - \left[ \frac{R_2 - B_2(T_c)}{R_{a2} - B_2(T_c)} \right]^{1/k'_{opt}(T_c)} \left\{ R_{a1} - f[B_2(T_c)] \right\} \right\} y_i$$
 (B.9)

Since Equation B.9 is a complicated function of  $T_c$ , it is impossible to obtain an estimate of  $T_c$  through analytical method. Instead, we select a series of values of  $T_c$  between 210 and 250K. We compute  $E^2(T_c)$  for each  $T_c$  and search for the  $T_c$  that is associated with the minimum value of  $E^2(T_c)$ . This  $T_c$  is our best first guess value of the mean cirrus cloud temperature.

# APPENDIX C. ITERATIVE MATCHING OF REFLECTANCE PAIRS

This section presents an iterative numerical method for the optimal matching of the computed and measured reflectance pair  $(0.645 - 1.6 \,\mu\text{m})$ .

Let the reflectances r ( $\lambda = 0.645 \, \mu m$ ) be  $r_1$ , and r ( $\lambda = 1.6 \, \mu m$ ) be  $r_2$ . Also, let a series of  $D_e$  and  $\tau$  values be chosen:  $D_{e1}, D_{e2}, ..., D_{en}, ..., D_{eN}$ , and  $\tau_1, \tau_2, ..., \tau_m, ..., \tau_M$ , where N and M are the total number of selected mean effective sizes and optical depths. We denote the subscripts 'comp' and 'meas' as the computed and measured values, respectively. Thus, radiance tables of  $r_{1comp}(n,m)$  and  $r_{2comp}(n,m)$  for all combinations of  $(D_{en}, \tau_m)$  are obtained from LBLE. We start by searching for the index  $m^{(0)}$ , so that:

$$r_1^{-(0)} < r_{1meas} \le r_1^{+(0)}$$
 (C.1)

where:

$$r_1^{-(0)} = r_{1comp}(3, m^{(0)})$$

and

$$r_1^{+(0)} = r_{1comn}(3, m^{(0)} + 1)$$

Once  $m^{(0)}$  is determined, we obtain  $\tau^{(0)}$  as:

$$\tau^{(0)} = \tau_{m^{(0)}} + \frac{\tau_{m^{(0)}+1} - \tau_{m^{(0)}}}{r_{n}^{+(0)} - r_{n}^{-(0)}} (r_{1meas,i} - r_{1}^{-(0)})$$
(C.2)

Then by using the index  $m^{(0)}$ , we search for the index  $n^{(0)}$ , so that:

$$r_2^{-(0)} < r_{2meas} \le r_2^{+(0)},$$
 (C.3)

where:

$$r_2^{-(0)} = r_{2comp}(n^{(0)} + 1, m^{(0)}); r_2^{+(0)} = r_{2comp}(n^{(0)}, m^{(0)} + 1)$$

With  $n^{(0)}$  determined, we obtain  $D_e^{(0)}$  as:

$$D_e^{(0)} = D_{e,n^{(0)}} + \frac{D_{e,n^{(0)}+1} - D_{e,n^{(0)}}}{r_2^{-(0)} - r_2^{+(0)}} (r_{2meas,i} - r_2^{+(0)})$$
(C.4)

We then go back and search for the index  $m^{(1)}$ , so that:

$$r_1^{-(1)} < r_{1meas} \le r_1^{+(1)}$$
 (C.5)

where:

$$r_{1}^{-(1)} = r_{1comp}(n^{(0)}, m^{(1)}) + \frac{r_{1comp}(n^{(0)} + 1, m^{(1)}) - r_{1comp}(n^{(0)}, m^{(1)})}{D_{e_{n^{(0)}+1}} - D_{e_{n^{(0)}}}} [D_{e}^{(0)} - D_{e, n^{(0)}}]$$

$$r_{1}^{+(1)} = r_{1comp}(n^{(0)}, m^{(1)} + 1) + \frac{r_{1comp}(n^{(0)} + 1, m^{(1)} + 1) - r_{1comp}(n^{(0)}, m^{(1)} + 1)}{D_{e, n^{(0)} + 1} - D_{e, n^{(0)}}} [D_{e}^{(0)} - D_{e, n^{(0)}}]$$

Once  $m^{(1)}$  is determined, we obtain  $\tau^{(1)}$  as:

$$\tau^{(1)} = \tau_{m^{(1)}} + \frac{\tau_{m^{(1)}+1} - \tau_{m^{(1)}}}{r_1^{+(1)} - r_1^{-(1)}} (r_{1meas,i} - r_1^{-(1)})$$
(C.6)

Then by using the index  $m^{(1)}$ , we search for the index  $n^{(1)}$ , so that:

$$r_2^{-(1)} < r_{2meas} \le r_2^{+(1)}$$
 (C.7)

where:

$$r_{2}^{-(1)} = r_{2comp} \left( n^{(1)} + 1, m^{(1)} \right) + \frac{r_{2comp} \left( n^{(1)} + 1, m^{(1)} + 1 \right) - r_{2comp} \left( n^{(1)} + 1, m^{(1)} \right)}{\tau_{m^{(1)} + 1} - \tau_{m^{(1)}}} \left[ \tau^{(1)} - \tau_{m^{(1)}} \right]$$

$$r_{2}^{+(1)} = r_{2comp}(n^{(0)}, m^{(1)} + 1) + \frac{r_{2comp}(n^{(0)} + 1, m^{(1)} + 1) - r_{2comp}(n^{(0)}, m^{(1)} + 1)}{\tau_{m^{(1)} + 1} - \tau_{m^{(1)}}} [\tau^{(1)} - \tau_{m^{(1)}}]$$

With  $n^{(1)}$  determined, we obtain  $D_e^{(1)}$  as:

$$D_{e}^{(1)} = D_{e,n^{(1)}} + \frac{D_{e,n^{(1)}+1} - D_{e,n^{(1)}}}{r_{2}^{-(1)} - r_{2}^{+(1)}} (r_{2meas,i} - r_{2}^{+(1)})$$
(C.8)

The iterative process continues until a minimum value for the following residual is reached:

$$E^{2}(r_{1}, r_{2}) = \{\ln[r_{1comp}(\tau^{(j)}, D_{e}^{(j)})] - \ln[r_{1meas,i}]\}^{2} + \{\ln[r_{2comp}(\tau^{(j)}, D_{e}^{(j)})] - \ln[r_{2meas,i}]\}^{2},$$
(C.9)

where  $r_{1\text{comp}}(\tau^{(j)}, D_e^{(j)})$  and  $r_{2\text{comp}}(\tau^{(j)}, D_e^{(j)})$  are interpolated values of  $r_{1\text{comp}}$  and  $r_{2\text{comp}}$  for the jth iteration value of  $\tau$  and  $D_e$ , respectively.

# APPENDIX D. ESTIMATING MEAN WATER CLOUD TEMPERATURE

This section describes an optimization scheme to find a first guess of mean water cloud temperature in the IR algorithm.

Using the VIIRS 3.7 and 10.8  $\mu$ m channel data, we proceed to solve the complete set of retrieval equations: 64 through 65, 66, and 70. The solution method follows that described in Ou *et al.* (1993, 1995, 1998). According to the flow chart shown in Figure 20, we need to make a first guess of  $\tau$ ,  $k_2/k_1$ , and  $r_e$ . For this purpose, we use an optimization method to obtain the best-fit mean cirrus cloud temperature. Subsequently, the first guess of  $\tau$ ,  $k_2/k_1$ , and  $r_e$  may be obtained from Equations 76 through 78. The optimization scheme is divided into two parts. The first part is to find an optimal value of  $k_2/k_1$ , given a cloud temperature  $T_c$ . The second part is to find an optimal value of  $T_c$ . The optimization scheme is described as follows.

In the first part, we rewrite Equation 70 in the form:

$$y = k' x \tag{D.1}$$

where:

$$y = \ln\left[\frac{R'_2 - B_2(T_c)}{R'_{\alpha 2} - B_2(T_c)}\right]$$
 (D.2)

$$x = \ln\left[\frac{R'_{1} - f[B_{2}(T_{c})] - (R'_{a1} - R'_{1})/\alpha_{1}}{R'_{a1} - f[B_{2}(T_{c})]}\right]$$
(D.3)

$$\mathbf{k'} = \mathbf{k}_{2} / \mathbf{k}_{1} \tag{D.4}$$

For a given cloud temperature  $T_c$ , we search for an optimal  $r_e$  that minimizes the residual:

$$E_{1}^{2}(k') = \sum_{i} (y_{i} - k' x_{i})^{2} y_{i}$$
 (D.5)

where i represents the pixel index. Equation D.5 is a weighted residual expression with weight  $y_i$ . Because for  $R_{2i} \rightarrow B_2(T_c)$ ,  $y_i \rightarrow -\infty$ , and for  $R_{2i} \rightarrow R_{a2}$ ,  $y_i \rightarrow 0$ , more weight is put on those pixels whose radiance is closer to the Planck function at  $T_c$ . Since Eq. D.5 is a nonlinear algebraic equation, we can only use numerical search method to find the optimal  $r_e$ . Using the parameterized functional form of k' and  $\alpha_1$ , we search for the minimal  $E^2_l(k')$  associated with each  $T_c$ .

In the second part, we again rewrite Equation 69 in the following form:

$$R'_{1}(1+1/\alpha_{1}) = f[B_{2}(T_{c}) + R'_{a}/\alpha_{1}] + \left[\frac{R'_{2} - B_{2}(T_{c})}{R'_{a2} - B_{2}(T_{c})}\right]^{1/k'_{opt}(T_{c})} \{R'_{a1} - f[B_{2}(T_{c})]\}$$
(D.6)

To obtain an estimate of mean cirrus cloud temperature  $T_c$ , we will minimize the following weighted residual:

$$E_{2}^{2}(T_{c}) = \sum_{i} \left\{ R'_{1} - f[B_{2}(T_{c})] - (R'_{a1} - R'_{1}) / \alpha_{1} - \left[ \frac{R'_{2} - B_{2}(T_{c})}{R'_{a2} - B_{2}(T_{c})} \right]^{1/k'_{opt}(T_{c})} \left\{ R'_{a1} - f[B_{2}(T_{c})] \right\}^{2} y_{i}$$
(D.7)

Since Equation D.7 is a complicated function of  $T_c$ , it is impossible to obtain an estimate of  $T_c$  through analytical method. Instead, we select a series of values of  $T_c$  between 210 and 250K. We compute  $E^2(T_c)$  for each  $T_c$  and search for the  $T_c$  that is associated with the minimum value of  $E^2(T_c)$ . This  $T_c$  is our best first guess value of the mean water cloud temperature.